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DOLLAR ILLIQUIDITY AND CENTRAL BANK SWAP ARRANGEMENTS DURING
THE GLOBAL FINANCIAL CRISIS

Andrew K. Rose
Mark M. Spiegel

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ABSTRACT

While the global financial crisis was centered in the United States, it led to a surprising appreciation in the dollar, suggesting global dollar illiquidity. In response, the Federal Reserve partnered with other central banks to inject dollars into the international financial system. Empirical studies of the success of these efforts have yielded mixed results, in part because their timing is likely to be endogenous. In this paper, we examine the cross-sectional impact of these interventions. Theory consistent with dollar appreciation in the crisis suggests that their impact should be greater for countries that have greater exposure to the United States through trade and financial channels, less transparent holdings of dollar assets, and greater illiquidity difficulties. We examine these predictions for observed cross-sectional changes in CDS spreads, using a new proxy for innovations in perceived changes in sovereign risk based upon Google-search data. We find robust evidence that auctions of dollar assets by foreign central banks disproportionately benefited countries that were more exposed to the United States through either trade linkages or asset exposure. We obtain weaker results for differences in asset transparency or illiquidity. However, several of the important announcements concerning the international swap programs disproportionately benefited countries exhibiting greater asset opaqueness.

Andrew K. Rose
Haas School of Business Administration
University of California, Berkeley
Berkeley, CA 94720-1900
and NBER
arose@haas.berkeley.edu

Mark M. Spiegel
Federal Reserve Bank of San Francisco
101 Market Street
San Francisco, CA 94105
mark.spiegel@sf.frb.org

An online appendix is available at:
<http://www.nber.org/data-appendix/w17359>

1 Introduction

Although the recent crisis exposed fragilities throughout the global economy, there can be no doubt that it originated and was centered in the United States. When difficulties arose in sub-prime mortgages in early 2007, investors became concerned about a wide set of U.S. assets, resulting in fire sales and the failure or near-failure of a number of systemically important U.S. financial firms [Bernanke (2009)]. Between October 2007 and October 2008, there was a \$8 trillion sell off in U.S. equity values [Brunnermeier (2009)]. A surprising feature of the recent financial crisis is that at its peak the American dollar actually rose in value. Going into the crisis, most thought that the adjustment process to undo the large global imbalances that had built up during the boom would include a sharp dollar depreciation [e.g. Krugman (2007)].

Instead, the crisis country currency appreciated [Engel (2009)]. For example, see Figure 1, which plots the VIX and VSTOXX measures of US and European equity market volatility respectively against the dollar-euro exchange rate during late 2008. The dollar exchange rate moved quite closely with volatility in equity markets, as can be seen by examining plots of the VIX and VSTOXX indices. This leads us to the view that the appreciation of the dollar resulted from a flight to liquidity rather than solely a flight to safety. While there probably was some movement towards safety [e.g. Fratzscher (2009), McCauley and McGuire (2009)], we concentrate on the liquidity issue here. Many studies [e.g. Baba and Packer (2009b)] characterize the illiquidity as a shortage in dollar funding suffered by financial institutions. Viewed from the prism of a global dollar liquidity shortage due to the unique role played by the dollar in global financial markets, the temporary appreciation of the dollar is unsurprising.¹

At the height of the crisis, the Federal Reserve extended dollar assets to major industrial

¹Goldberg and Tille (2008) show that the dollar plays a prominent role in invoicing in international transactions, even in many that do not involve agents from the United States. Similar concerns drive currency invoicing decisions in debt issuance [Chinn and Frankel (2007)]. The impact of scale effects has been demonstrated in the case of the advent of the euro, where the increased volume of existing issuance in euro relative to national currencies resulted in a substantial move towards the euro in new issuance [Hale and Spiegel (2008)].

countries, and several emerging markets' central banks to alleviate these dollar shortages.² Obstfeld, Shambaugh, and Taylor (2009) note that desirable alternatives to the swap arrangements did not exist, as increased domestic currency extensions from local central banks could have led to undesirable currency depreciation, and the use of foreign central bank dollar reserves would have reduced their holdings, raising anxiety.³ They argue that the broad injection of dollar liquidity was "... one of the most notable examples of central bank cooperation in history ..."

The evidence on the impact of central bank interventions is mixed. Some of the studies [e.g. Taylor and Williams (2009)] find no impact, while others, such as McAndrews, Sarkar, and Wang (2008), find significant but small impacts. More recent studies, such as Baba and Packer (2009b), concentrate on the most turbulent portion of the crisis and find larger effects. However, the endogeneity of these injections, which were provided when and where they were most needed, poses a challenge in evaluating their impact.

Given these difficulties, we examine the cross-sectional impacts of central bank efforts to address dollar-funding shortages. We begin with a descriptive overview of the central bank responses to the global financial crisis, reviewing a number of the relevant empirical regularities that have been found in the literature. We then discuss the implications of a theoretical model derived in a companion paper [Rose and Spiegel (2011)] – and summarized in the appendix – that describes the crisis as stemming from toxic American assets but still predicts the observed dollar appreciation.

We then bring the cross-sectional predictions of that model to the data to reassess the impact of the attempts by the Federal Reserve and others to inject dollar liquidity into the global financial system. Theory suggests that the impact of these injections should be greater among countries that have greater exposure to the United States through trade and financial channels, less transparent holdings of dollar assets, and greater illiquidity difficulties. We test these hypotheses by examining

²Some have also suggested that the swaps were motivated by a desire to mitigate the aforementioned exchange rate pressures.

³Some emerging market country swap arrangements reflected their desire to avoid obtaining funds from the International Monetary Fund, and may have more reflected the need for hard currency reserves [e.g. Engel (2009)].

the impact of announced U.S. dollar auctions by foreign central banks, weighted by the size and average maturity of auctioned assets, on CDS spreads for a large cross-section of countries. We find robust evidence that the auctions disproportionately benefited countries that were more exposed to the United States, either through trade or financial channels, as the theory predicts. We obtain weaker or incorrect results for national differences in the impact of the auctions by the transparency of their dollar holdings and measures of illiquidity.

We also examine the impacts of the major announcements concerning the international swap arrangements. For several of the most important announcements, such as the one that removed the ceilings on swaps with major foreign central bank partners and the announcement initiating swap arrangements with a broader set of countries, our results for announcements roughly match those for the actual auctions. However, for others, such as the actual launch of the program, we find disproportionate benefits among countries exhibiting greater illiquidity.

The following section reviews the evidence in the literature on the impact of the central bank swap lines on global financial conditions. Section 3 discusses our base empirical specification. Section 4 subjects our results to a battery of robustness tests. Lastly, section 5 concludes.

2 Evidence on the impact of the swap arrangements

Major announcements concerning international swap lines by the Federal Reserve during this period are shown in Table 1. The first is December 12, 2007, when the Federal Reserve announced its swap arrangements with the European Central Bank (ECB) and the Swiss National Bank (SNB). These were initially capped at \$20 and \$4 billion respectively. With the increased turmoil in global financial markets in the fall of 2008, swap lines were extended and expanded. On September 18, 2008, lines were introduced for the Bank of England (BOE), the Bank of Japan (BOJ) and the Bank of Canada, while lines with the ECB and the SNB were increased. Less than a week later,

on September 24, swap facilities were introduced for the Reserve Bank of Australia, the Swedish Riksbank, the Denmark National Bank, and the Norwegian Central Bank. In October of the same year, existing lines were "uncapped," on October 13 for the BOE, the ECB and the SNB, and on October 14 for the BOJ. Finally, on October 28, 2008, lines were introduced for New Zealand, and on October 29, for Brazil, Mexico, Korea, and Singapore.⁴ The range of swap lines was also broadened over this period from longer-term offers (one to three months) to also include one week and overnight offers, and from primarily repos and collateralized loans to also include foreign exchange swaps [Ho and Michaud (2008)]. Other nations, including the Swiss National Bank and the ECB, also entered into swap arrangements with other countries with funding needs in those countries' currencies.

These swap lines allowed these foreign central banks to access dollar-denominated assets which they could then lend to their financial institutions that were experiencing dollar illiquidity. At the height of the program at the end of 2008, draw downs reached \$291 billion at the ECB, \$122 billion at the BOJ, and \$45 billion at the Bank of England [Goldberg, Kennedy, and Miu (2010)].

Other central bank efforts to inject dollar liquidity were also initiated. The term auction facility (TAF) program, aimed at providing funds to financial institutions, was introduced in December of 2007. Through this facility, depository institutions were able to borrow directly from the Federal Reserve without using the discount window [Taylor and Williams (2009)].⁵ The ECB also conducted dollar term funding auctions. These were supported by the swap lines with the Federal Reserve and provided dollar funds to institutions in the European Union with ECB-eligible collateral [Baba and Packer (2009a)] (See Figure 2).

The volume of TAF auctions increased dramatically during the fall of 2008, coinciding with the

⁴See Ho and Michaud (2008) and Goldberg, Kennedy, and Miu (2010) for reviews of the details of the central bank swap programs during the crisis.

⁵As Taylor and Williams (2009) point out, it is important to remember that the liquidity effects of the TAF auctions are not due to any increase in total bank reserves of the amount of "high-powered money" in the financial system, as bank borrowing was offset by open market sales of securities.

dates of the Lehman failure and the subsequent market turmoil. As financial conditions improved, however, the terms offered under the overseas swap facilities became less desirable. Offer rates for dollar swap facility funds reached about 100 basis points higher than terms available to US and some foreign financial institutions under the TAF program. Moreover, by the first quarter of 2009 the market terms had improved to the point that participation in central bank swaps would only have been attractive to institutions lacking access to funds in private markets or lacking collateral necessary to participate in the TAF program [Goldberg, Kennedy, and Miu (2010)]. The volume of draw-downs decreased quickly as conditions improved.

The swap arrangements were a crucial part of efforts by global officials to restore liquidity to the financial system, as evidenced by the enormous draw downs at the end of 2008. Table 2 reports the volume and average tenor of the TAF funds auctioned by the four major central banks over the course of our sample. It is clear that this injection of dollar-denominated capital was large, with volumes peaking in the fall of 2008 for the four major foreign central banks.⁶

A number of studies have emerged attempting to gauge the success of the programs in improving global dollar liquidity. In an early study, Taylor and Williams (2009) examine the impact of the TAF auctions. They find no impact of these auctions on the 3-month spread of unsecured LIBOR lending rates over overnight index swaps (OIS), which they take as a proxy for interest rate expectations. Their work was followed by a number of researchers, including McAndrews, Sarkar, and Wang (2008), who argued that a proper assessment of the impact of the TAF auctions required looking only at changes in the LIBOR-OIS spreads on days of announcements and auction operations. Using this methodology, they find that the TAF auctions and announcements accounted a cumulative reduction of more than 50 basis points in the OIS-LIBOR spread. Moreover, they find that international TAF auctions also had a statistically significant and even larger impact on

⁶The popularity of the swap arrangements imply some market failure in international financial markets, particularly among the major central banks who under normal circumstances would likely be able to raise adequate funds on their own.

spreads than domestic auctions. Both McAndrews, Sarkar, and Wang (2008) and subsequent work by Taylor and Williams (2008) based on spreads find that announcements had larger impacts than actual auctions.

Other efforts to characterize the impact of the central bank dollar injections concentrate on evidence from the FX swap market. As discussed in Baba and Packer (2009b), disruptions in the FX swap market began appearing at the height of the financial crisis. FX swap prices began to reflect increases in perceived counter-party risk among European financial institutions, as doubts grew about the abilities of these institutions to fulfill their dollar obligations. This resulted in deviations from short-term covered interest parity. Baba and Packer (2009b) find that the establishment of the international fund lines, as well as the dollar term funding auctions financed by these swaps, had a significant downward impact on observed deviations from covered interest parity in the FX swap market. They obtain mixed results, as US dollar auctions are found to have had a robust negative impact on deviations to covered interest parity subsequent to the Lehman failure, but not before. Similar results are reported in Baba and Packer (2009a) and Hui, Genberg, and Chung (2010).

The impact of the central bank actions on a broader set of countries is examined by Aizenman and Pasricha (2010). They concentrate on emerging market economies that were granted swap arrangements by the Federal Reserve at the height of the crisis. They demonstrate that the set of emerging market economies that received swap arrangements were selected in part on the basis of having exceptionally large outstanding obligations to the Federal Reserve. Their results indicate that the establishment of swap arrangements had little impact on national credit default swap spreads, but did contribute to exchange rate appreciation, or at least stemmed exchange rate depreciation.⁷

⁷ More recently, there have also been efforts to assess the impact of the large scale asset purchase (LSAP) program conducted by the Federal Reserve. These studies, including Hamilton and Wu (2011), Krishnamurthy and Vissing-Jorgensen (2011), and D'Amico and King (2011) all find substantial impacts of the LSAP programs had substantial impacts on interest rates.

Overall, it is safe to characterize the evidence on the impact of central bank interventions as mixed. Even the work of McAndrews, Sarkar, and Wang (2008), which was subsequently confirmed by Taylor and Williams (2008), only finds about a 2 basis point impact of TAF events on LIBOR-OIS spreads. While it may not be surprising that the dollar auctions had their greatest effect during the height of the turmoil, it is safe to say that the magnitude of the observed responses during the pre-Lehman period was disappointing. Indeed, it was during this period unprecedented policies were adopted, providing a reminder that while this period was not turbulent relative to what immediately followed, it was still exceptional relative to recent historical data.

A number of difficulties have been pointed out with time series-based evidence. One problem is that these approaches implicitly ascribe all movements not covered by measured changes in counter-party risk to the policy action, while a substantial number of other developments were simultaneously taking place [Taylor and Williams (2009)]. Another is that there is clear evidence that central bank swap policies have been endogenous: Central bank swap partners were clearly not chosen at random. Moreover, Aizenman and Pasricha (2010) find that the set of emerging market economies chosen as candidates for swap arrangements are notable in the magnitude of their outstanding US debt obligations. In addition, the timing of the largest interventions exactly coincides with the period of greatest turmoil. Finally, one would think that private agents would consider an announcement concerning the design of the international swap program as revealing something about the central banks' views about the severity of the crisis situation. The time series evidence has difficulty separating the direct impact of the program from its impact through private sector expectations.⁸

⁸One notable exception is D'Amico and King (2011) who identify significant impacts of the LSAP programs in a cross-section of securities. Moreover, they identify effects of pre-announced asset purchases, which they term "flow effects," which are related to the pre-announced injections of dollar liquidity on auction dates that we study below.

3 Empirical specification

Given the problems discussed in the previous section with existing methodologies, along with the mixed results in the literature, our empirical strategy is to identify cross-sectional restrictions that can be taken to the data to identify the impact of the central bank actions. This approach avoids the timing and endogeneity issues associated with the event-study approaches in the literature. In this section, we first review theoretical underpinnings motivating heterogeneity in the expected impact of the auctions. We then introduce our data set and present basic results.

3.1 Theoretical motivation

It seems natural to turn to the literature on money demand based on microeconomic frictions to examine the role of dollar illiquidity in the surprising dollar appreciation during the recent crisis. Early studies, such as Kiyotaki and Wright (1993) and Trejos and Wright (1995) established that a role for money that leads to positive money demand can be motivated within a search model where money acts as a convenient medium of exchange due to its superior liquidity, avoiding the need for a double coincidence of wants. More recently, Lagos and Wright (2005) develop a tractable search-based monetary model by dividing each period into two sub-periods: In the first, agents enter a centralized market in which all goods and assets clear in a very standard manner. However, agents then move on to a decentralized market with anonymous bilateral matching and a double-coincidence problem. The combination of these two markets allows for the incorporation of bargaining under interesting conditions, including the possibility of illiquidity, with tractability ensured by the fact that the next period all agents reunite in the centralized market, where outcomes are degenerate and in particular do not depend on the distribution of money holdings across agents. This methodology was extended further in Lester, Postlewaite, and Wright (2009b), who develop a closed-economy model where assets differ in their general acceptability, and hence liquidity. In their model, assets may be of high or low quality, and agents that are uninformed refuse to accept

low quality assets in exchange.⁹

In a companion paper [Rose and Spiegel (2011)], whose details are summarized in the appendix of this paper, we develop an international version of the search-based asset model of Lagos and Wright (2005).¹⁰ In this model, assets differ in their returns, their "opaqueness," and in their liquidity. The possibility of illiquidity arises because, as in Lester, Postlewaite, and Wright (2009b), agents trading in decentralized markets reject opaque assets whose value they don't recognize. We demonstrate that a decline in the yield on the opaque US asset decreases the stock of dollar assets available for transactions purposes, and raises demand for other US assets, such as currency, thereby resulting in an appreciation of the dollar exchange rate. Broadly, we interpret the decline in the yield on the real asset as analogous to the fall in the perceived value of exotic US assets during the global financial crisis, and the appreciation of the dollar relative to the value of the other national currency as analogous to an increase in the relative yield of safe US assets.¹¹

This model has implications for the predicted impact of the central bank auctions conducted with dollar funds obtained from the Federal Reserve. We consider the capital injections under the swap program as analogous to an increase in the stock of dollar assets held by agents on entering the market that exhibits dollar illiquidity.¹² In the appendix, we demonstrate that the benefits of this injection are increasing in three characteristics: The first is the probability of needing to transact in US dollars in the decentralized market, which we proxy with alternative measures of exposure, as agents with greater exposure to the United States are more likely to find themselves

⁹See Lester, Postlewaite, and Wright (2009a) for a demonstration that equilibria in which agents reject assets that they do not recognize at any price are feasible.

¹⁰Geromichalos and Simonovska (2010) and Liu (2010) also develop international versions of the Lagos and Wright model. We also include full proofs of the results used in this paper in a technical appendix posted online at <http://www.frbsf.org/economics/economists/mspiegel/wp11-18appendix.pdf>.

¹¹We do not want to suggest that this channel was the only source of dollar illiquidity. Brunnermeier (2009) discusses the "liquidity spirals" that resulted from declines in asset prices that deteriorated bank balance sheet positions, leading to further tightening of lending standards. Emerging market countries also had a need for foreign currency reserves.

¹²One additional benefit of the swap programs that is outside of our formal model was that financial institutions were not given cash, but US Treasury securities that are commonly admissible as collateral.

in need of dollars for transactions or servicing liabilities. The second is the probability of being paired with an uninformed agent, which we interpret as reflected in the "opaqueness" of a country's aggregate dollar holdings. Finally, the impact is predicted to be increasing in the severity of dollar illiquidity in the country.¹³

3.2 Base Specification

Our base specification includes examines the cross-sectional restrictions implied by the theory. Initially, we look at an event study specification by examining the average implications of the TAF auctions across the sample, measured by an event dummy corresponding to the week of the auctions, along with interactive slope variables to capture the extra sensitivity exhibited by countries of certain characteristics suggested theoretically. In addition, we include a number of conditioning variables. Our initial specification is¹⁴

$$\begin{aligned} \Delta CDS_{it} = & \alpha_i + \theta_t + \beta_1 Exposure_{it} \cdot SP500_t + \beta_2 Exposure_{it} \cdot auction_{t-1} \\ & + \beta_3 Transp_{it} \cdot auction_{t-1} + \beta_4 Illiquid_{it} \cdot auction_{t-1} + \beta_5 \Delta Default_{it} + \epsilon_{it}. \end{aligned}$$

where ΔCDS_{it} represents the change in CDS spreads on country i sovereign debt during week t ; α_i is a country dummy; $Exposure_{it}$ represents exposure to the United States, measured as discussed above; $SP500_t$ represents the annualized percentage change in the S&P 500. $auction_{t-1}$ is equal

¹³Peter and McGuire (2009) also argue that exposure matters, arguing that differences in financial system balance sheet exposure to US assets are likely to be positively correlated with dollar shortage vulnerabilities. While our model literally looks at liquidity shortages in trade, we also consider financial exposure to the United States, such as the exposure measures in Rose and Spiegel (2009b).

¹⁴We examined two additional specifications. First, we conducted an event study specification examining the average implications of the TAF auctions across the sample along with the interactive slope and conditioning variables used in our base specification. Here the results were disappointing, as can be seen in Table A1, which mirror the weak event study results in the literature. The interacted exposure variable consistently obtains its expected negative coefficient value, but is usually insignificant. Moreover, the $auction_{t-1}$ variable consistently enters with the wrong positive sign, although it is almost always insignificant. Second, we added the variables of interest on their own, i.e. not interacted with the TAF volume and tenor. The results were much the same as those reported for our base specification, and are shown in Table A2.

to the sum of the volume of each auction during week $t - 1$ times the average tenor of that auction in weeks where auctions took place, and 0 in weeks with no auctions.¹⁵ $Transp_{it}$ represents dollar asset transparency, measured as the ratio of dollar equity holdings to the sum of holdings of dollar equities plus short and long-term agency debt; $Illiquid_{it}$ represents asset illiquidity, measured as the ratio of short-term US liabilities to total exports; $\Delta Default_{it}$ conditions for changes in perceived default risk, based on our proxy from Google search, discussed below; and ϵ_{it} is a disturbance term, assumed to be well behaved.

Our three variables of interest are the interactive terms representing the relative impact of the auctions on country i dollar liquidity by exposure, asset transparency, and illiquidity: $Exposure_{it} \cdot auction_{t-1}$, $Transp_{it} \cdot auction_{t-1}$, and $Illiquid_{it} \cdot auction_{t-1}$.

The remainder of the variables are nuisance terms meant to capture other potential determinants of movements in sovereign CDS spreads, including $Exposure_{it} \cdot SP500_t$, which is meant to pick up the impact on country i of other economic developments in the US, $auction_t$, which is meant to pick up the average impact of the TAF auctions across countries, and $\Delta Default_{it}$ which is our Google measure meant to capture changes in the public's perception of default risk in country i . α_i and θ_t represent country and time dummies respectively.¹⁶ The time fixed effects address a number of issues: the foreign TAF auctions were just one component of a number of policy responses by the Federal Reserve, as well as both the US Treasury, and Treasuries and central banks around the world. In addition, the composition of borrowers and the size and tenor of swap arrangements varied over the course of the policy as the swap programs were expanded. However, these fixed effects would be collinear with the $auction_{t-1}$ variable, as the timing, total volume and average tenor of

¹⁵We use lagged weeks for the auction variable because many auctions took place late in the week, requiring some time for the market response in terms of the impact on other nations to be felt. Recall that these auction events have all been previously announced, and hence are not surprises.

¹⁶While our specification is of weekly frequency, we only use monthly time dummies in the above specification. When we use weekly time dummies, all of the variables, including both our variables of interest and the nuisance parameters are very insignificant, as can be seen in Table A4 in the appendix.

auctions are common across countries.¹⁷

3.3 Data

3.3.1 Standard data

Our full sample is based on weekly data, and runs from December 10, 2007 to December 31, 2009. Our sample is a broad panel of emerging market and smaller developed economies, and includes 30 OECD and 38 non-OECD countries.

We consider two types of measures of "exposure" to the United States. First, we consider trade-related measures, such as *Exports*, *Imports* and total *Trade* with the United States, as a share of total global trade. These variables are closer to the explicit model above, in the sense that we would expect that agents with more trade with the United States would be more likely to find themselves with potentially profitable trade opportunities with US nationals. We use monthly data on trade exposure to the United States from the IMF Direction of Trade statistics.

We also consider a variety of measures of asset exposure, including *Assets(TIC)*, which measures total holding of US assets based on TIC data as a share of global assets measured using the IMF CPIS data set. We also consider two subsets of this data, *Debt*, and *LTDebt*, which measure total claims on US debt and total claims on US long-term debt respectively. Both numerator and denominator of these variables are available only annually.¹⁸ *Assets(CPIS)* represents an annual proxy for US asset exposure as a share of total global asset holdings, according to the IMF CPIS data set.¹⁹ Estimation is done by OLS using robust standard errors clustered by country.

¹⁷We examine the possibility of extra sensitivity in the countries directly receiving the auction funds below.

¹⁸The TIC data is measured is data is annual, based on exposure in June, while the CPIS data is annual, based on December exposure. We use TIC data for a given year as a proxy for exposure from July of the previous year to June of the current year, and use CPIS data for a given year as a measure of exposure from January to December of that same year. Ratios are then constructed from these monthly series as global exposure is only available from the CPIS data set. This led to some calculated ratios for these variables having implausible values, either less than 0 or greater than 1. In response, we censor these variables to have minimum value 0.01 and maximum value 1.

¹⁹Below, we report results based on trade and asset exposure as separate specifications. However, we also ran

Data on foreign central bank auctions was obtained from the Federal Reserve Board of Governors, as were the details of announcements concerning changes in the Federal Reserve’s swap program. We condition auction ”events” for two characteristics: volume in overall dollar value and average tenor in days of length of contracts auctioned. The latter adjustment is important because securities auctioned varied from high maturities of 95 days to maturities as low as one day, representing substantially different levels of effective liquidity injections per dollar issued (see Table 2 for summary statistics for major central banks).

We obtain weekly percentage changes in CDS spreads and S&P500 returns from Bloomberg.

3.3.2 Default risk proxy from Google search

Our primary non-standard data series is a proxy for perceived changes in country creditworthiness. Since LIBOR rates are limited to a small set of developed nations, we follow Aizenman and Pasricha (2010) in using differences in CDS spreads as our indicator of liquidity risk.²⁰ Of course, changes in country creditworthiness will also affect CDS spreads, so we need to condition on country creditworthiness in order to isolate the movements in CDS spreads attributable to liquidity changes. This is problematic for the broad cross section that we use in our study, as many of the countries in our sample do not have widely-traded instruments that one might typically consider as potential indicators of changes in a country’s creditworthiness. Aizenman and Pasricha (2010) use Economist Intelligence Unit data for their sample of emerging market economies, but such data is only available monthly.

In response, we use weekly search data obtained from Google Insights for Search. Based

specifications with a form of both types of exposure included, and obtained similar results. We also investigate a number of alternative exposure measures. First, we normalize exposure by country GDP instead of global exposure. Second, we account for the fact that exposure to Europe is likely to be poorly measured because European assets are often held in tax havens in other countries in two ways: we look at bank exposure to the US, which is available consistently for all countries, and we also aggregate across the euro area. Our results are largely robust to all of these alternative exposure measures, as shown in appendices A10, A11, and A12.

²⁰Aizenman and Pasricha (2010) provide a theoretical model that links liquidity with CDS spreads.

on their own description [e.g. Google (2011)], Google Insights for Search analyzes a portion of worldwide Google web searches from all Google domains to compute how many searches have been done for a chosen group of terms relative to the total number of searches done on Google over time.

Google search data has been used in a number of studies. Choi and Varian (2009) use search data results to predict levels of economic activity for automobile sales and unemployment figures. Mondria, Wu, and Zhang (2010) find that increased search volume on Google is associated with greater inward investment and Da, Engelberg, and Gao (2011) demonstrate that increased activity is associated with temporary increases in equity values. In both of these studies, the effect is attributed to increased "attention."

Such real-time data is most often used to describe current economic conditions, rather than forecast future ones, in a growing application commonly referred to as "nowcasting." Studies have verified a number of cases where the Google search data have added information over and above that available from other sources [e.g. Varian (2010) and Kholodin, Podstawski, and Siliverstovs (2010)].

This is the sense in which we use the Google search data in our study. To measure changes in the perceived sovereign risk of a country, we use the relative incidence of searches of words related to default risk combined with that country's name. The percentage change in search volume for a given country combined with these default-related terms is then used as a proxy for changes in concerns about default risk about that country.²¹

A number of features of Google Insights should be pointed out. Responses are reported on a scale of 0 through 100. Figures are scaled by the highest volume response, which is given score 100. Remaining figures are then scored as their values as a share of the top reported value. Google

²¹We freely acknowledge that our Google data might be a better gauge of popular concern about a particular country's default risk, rather than that held by market professionals, as they more likely use propriety sources of data. Still, our results below demonstrate that there is a correlation with sovereign credit ratings, which presumably reflect the opinions of market participants rather than the general public.

also normalizes its series by a common variable, so values represent likelihoods of searches for a given country, rather than the absolute number of searches. This leaves all series country-specific. However, these series suit our purposes because we are only interested in the changes in our series over time, and the normalizations drop out.

One potential problem with our use of Google Insights as a proxy for changes in perceived default risk is that for proprietary reasons Google does not provide numerical values for responses when they fall below a certain threshold. For our purposes here, we proxy the numerical value for such observations as equivalent to the lowest reported value, which is clearly an upper-bound estimate of its true value.

To increase the potential correlation between our proxy and actual perceptions of creditworthiness, we choose a set of credit-risk related search words that are correlated with observed changes in perceived creditworthiness. Obviously, other estimates of changes in perceived country risk are not available at the high frequency that we use in our cross-section panel; this is what drove us to use the Google search data in the first place. We therefore examine the validity of our proxy by determining its correlation with other measures of default risk at the lower frequencies at which those other measures are available.

We begin with a set of 33 default-related words. While it would be desirable to evaluate all of the possible combinations of these words, this methodology is not possible because of restrictions placed by Google on the total number of searches that can be conducted on a single day.²²

In response, we developed a simple algorithm to choose the set of default-related terms we use to conduct the Google searches. First, we generated a full set of searches with each of the countries in our sample and one of the 33 default-related terms. We then regressed panels of these combinations of searches by countries and single default-related terms on monthly changes in Fitch sovereign ratings. We examine three series, foreign and domestic long term debt obligations and

²²We search over 112 countries for every variety of default-related terms.

short-term foreign obligations. Of these, we were most interested in the results for foreign long-term obligations.

Our results for foreign long-term obligations are shown in Table A3.²³ We found three words which entered significantly for all of the Fitch series: crisis, financial, and freeze.²⁴ We then ran searches with these three terms and one of the remaining words. This yielded six words which improved the fit of the Google searches with in-sample changes in Fitch ratings: "credit", "debt", "exposure", "liability", "recession", and "safety". We chose the set of four words that fit the best, which added the word "recession." We then examined the implications of adding a fifth word from this list. None of these improved the fit of our ratings changes regressions, so we settled on searches mentioning a country and one of four default-related terms: crisis, financial, freeze, and recession.

Correlations in the data between search volume and bond ratings changes are demonstrated in Figure 3. We plot the Google series for four countries, Iceland, Latvia, Greece and Ireland. Data availability differs by country, from as far back as 2004 for Ireland to 2008 for Iceland. However, all countries have data for the bulk of the crisis period. It can be seen that there is a lot of variability in the data, but all four countries appear to have credit downgrading episodes that correspond to local spikes in the Google series. Of course, there are lots of other spikes in the Google data that do not correspond to a credit downgrading event, and the relationship does not always appear to be exactly contemporaneous. Still, we would at least like to feel certain that changes in the Google ratings do correspond to changes in search volumes.

To investigate this question more formally, we considered the following panel specification for our entire cross-country sample:

²³Results for domestic long-term and foreign short-term are available on request.

²⁴We also found that the word "danger" entered significantly for short-term obligations, but neither of the other series. When adding this word to the 3 word base, however, the quality of fit deteriorated. In response, we continued with the 3 word base discussed in the text.

$$\% \Delta Rating_{it} = \alpha_t + \theta_i + \beta_1 \Delta Google_{it} + \epsilon_{it}. \quad (1)$$

where $\Delta Rating_{it}$ is the change in country i 's Fitch credit rating at time t , with one point for each change, α_t and θ_i are time and country dummy variables respectively, $\% \Delta Google_{it}$ is the variable of interest, the percentage change in the Google default proxy (hereafter referred to as default), and ϵ_{it} is an independent error term, assumed to be well-behaved.

Our results are shown in Table 3 for both the full time series over which Google search data is available and a smaller time series that corresponds to the period covered in our study below. Data is monthly, and our specification includes country and time fixed effects.

It can be seen that there is a strong negative relationship between ratings changes and Google search volumes in our full data panel, that is robust across the three different asset categories whose ratings we consider. The estimated coefficient values suggest that a doubling of Google search volume is predicted to, for example, result in a downgrade of foreign short-term debt equal to 6 basis points, even after controlling for changes in global conditions through the time fixed effects, for the time series corresponding to our study below.

We find it reassuring that the Google search volume data tracks this manifestation of changes in expectations about sovereign default risk in the manner we desire. We therefore use changes in the volume of Google index searches for a country name and one of the words associated with sovereign risk listed above as a proxy for changes in the public perception of default risk for that country.²⁵

²⁵As a robustness check, we also took an ad hoc set of default-related terms and used search results for that string instead of the stepwise procedure discussed above. These words included "risk", "default", "recession", "deficit", "debt", "crisis", and bankruptcy. Our reported results were robust to this alternative proxy, and are shown in Table A13.

3.4 Results

Our results are shown in Table 4. In terms of the three variables of interest, the interactive *Exposure* variable consistently obtains a negative sign, either for trade-related measures of exposure (Models 1 through 3), or for the measures of asset exposure (Models 4 through 7), with the exception of Model 5 which obtains the predicted negative sign, but is insignificant.

Moreover, the coefficient estimates suggest that discrepancies across countries with different exposure levels are substantial. Our dependent variable is measured in percentage changes in CDS spreads, which implies that the predicted decrease in CDS spreads from a week with average auction volume and tenor in our sample would be 36.5 basis points larger for a country with one standard deviation higher trade exposure to the United States as measured by our *Trade* variable. Similarly, the predicted decrease in CDS spreads from a week with average auction volume and tenor in our sample would be 26.2 basis points larger for a country with one standard deviation higher asset exposure to the United States, as measured by our *Assets(TIC)* variable.²⁶

The interactive *Transp* variable robustly enters significantly with its unpredicted negative sign. This suggests that this variable is likely picking up some benefit from having a relatively large stock of US Treasuries that allowed countries to fare disproportionately well on weeks with TAF auctions that is outside of our theoretical model. It may be that those countries whose public and private agents hold a transparent US dollar portfolio – measured in our data as the share of long and short-term US treasuries in total US holdings – have a greater need for dollar liquidity during crisis periods than those that do not. This need may be time-varying, and therefore not conditioned for by our country fixed effects.

Finally, the interactive *Illiquid* variable is insignificant throughout.

²⁶These calculations are based on the standard deviation of of the *Trade* exposure measure in our sample being 0.10, the mean values of weekly auction volume*tenor being 1.15, and the standard deviation of the *Assets(TIC)* variable being 0.41

Among our nuisance parameters, the $Exposure_{it} \cdot SP500_t$ variable is again significant with its predicted negative sign throughout, while the $\Delta Default_{it}$ obtains its predicted positive coefficient estimate, but is statistically insignificant throughout, with the exception of Model 5 which measures exposure as the share of U.S. asset holdings using the CPIS data.

We conclude that the foreign TAF auctions disproportionately benefited those countries more exposed to the United States, either through trade or asset exposure. However, we obtained exactly the wrong sign for the opacity of US asset holdings, suggesting that we pick up an effect not predicted by our theory. Finally, we obtained insignificant results for the interactive illiquidity variable.

4 Robustness Tests

In this section, we subject our chosen base specification to a number of robustness checks, including using alternative measures of illiquidity, alternative sub-samples of the data, and examining the impact of announcements concerning the international swap arrangements, rather than the auctions themselves.²⁷

²⁷We also conducted a number of other robustness tests which are reported as appendix tables. First, we considered changes in exchange rates, both as a potential additional independent variable, as they might represent an alternative driver of CDS spreads, and as a dependent variable, as changes in exchange rate pressure might be an alternative outcome of the auctions. Our results are reported in Tables A14 and A15 respectively. Our base regression results are robust to the inclusions of the exchange rate as an additional right hand side variable. However, we get far different results for the impact of the auctions on exchange rates. We find that illiquid countries experienced significantly greater relief in exchange rate pressure, usually at statistically significant levels, but the exposure variables are all insignificant. Of course, many things may drive exchange rate movements beyond the explicit model above and in patterns that are not well understood by economists. We also examined changes in LIBOR-OIS spreads as an alternative dependent variable. We have a much smaller sample, as we are limited to 8 countries. The results are shown in Table A16. We continue to obtain negative coefficient estimates throughout for exposure, but only at statistically significant levels in two of the 7 specifications. However, an additional specification is significant at a 10% confidence level.

4.1 Alternative Illiquidity Measures

We consider three alternative liquidity measures. These include short-term debt as a share of GDP, the ratio of short-term debt to international reserves, and the "Greenspan-Guidotti" illiquidity measure, which is measured as the ratio of short-term debt minus international reserves to international reserves.²⁸ Except for these alternative illiquidity measures, we keep our base specification and again consider all seven exposure measures used above.

Our results are shown in Table 5. For space considerations, we only report the results for the three interactive variables of interest.²⁹ We first measure illiquidity as the ratio of short-term debt to GDP. It can be seen that the results are qualitatively identical to those in our base specification. The interactive exposure variables are significantly negative throughout, with the exception of Model 5 with similar coefficient values. The interactive transparency variable again enters significantly with a negative sign throughout, while the interactive illiquidity variable is insignificant.

We next measure illiquidity in terms of the ratio of short-term debt to reserves. This specification again obtains a statistically significant negative sign for all of the interactive exposure variables except Model 5, negative and significant coefficient estimates on the interactive transparency variables, and insignificant coefficient estimates for the illiquidity measure.

Finally, we use the "Greenspan-Guidotti" measure of illiquidity, namely the ratio of short-term debt to reserves minus one. The interactive exposure measure again enters significantly with its expected negative sign for all of the trade-related exposure measures, but is significant for only one of the financial exposure variables, that of Model 5 which measures exposure as the ratio of holdings of US assets as a share of total global asset holdings.

The interactive transparency variable continues to obtain a negative coefficient, but is now

²⁸The latter two terms are similar, but the interaction with the volume and tenor variables imply that they are not identical, as shown in the results.

²⁹The full results are in appendix tables A6, A7, and A8.

insignificant throughout. The interactive illiquidity variable is again insignificant, except for the Model 4 specification that measures US exposure using the Assets TIC data.³⁰

While the financial exposure measures were a little weaker using the "Greenspan-Guidotti" measure of illiquidity, overall the results of the base regression reported above appear to be robust to the alternative illiquidity measures we entertained here.

4.2 Alternative Samples

We next consider dividing up our pooled sample into OECD and non-OECD sub-samples. It is quite plausible that these groups experienced different impact of the foreign TAF auctions. We again use our base specification with the seven different exposure measures.

The results for the OECD sub-sample are shown in Table 6. These results are quite similar to those in our base specifications, and stronger in some dimensions. The exposure variables all enter significantly with their expected negative signs including that of Model 5 this time. Moreover, the coefficient values are somewhat larger than those we obtained for the full sample. Moreover, the coefficient estimates suggest that discrepancies across countries with different exposure levels are again substantial.

For the OECD sub-sample, we find that the predicted decrease in CDS spreads from a week with average auction volume and tenor in our sample would be 44.8 basis points larger for a country with one standard deviation higher trade exposure to the United States as measured by our *Trade* variable. Similarly, the predicted decrease in CDS spreads from a week with average auction volume and tenor in our sample would be 33.1 basis points larger for a country with one standard deviation higher asset exposure to the United States, as measured by our *Assets(TIC)* variable.³¹

³⁰One problem with our liquidity measures is that Ireland is a major outlier. For example, Ireland's Greenspan-Guidotti illiquidity measure is standard deviations above the mean. Nevertheless, we obtained similar results throughout after dropping Ireland.

³¹These calculations are based on the standard deviation of the *Trade* exposure measure in our sample being 0.08, the mean values of weekly auction volume*tenor being 1.15, and the standard deviation of the *Assets(TIC)*

Among the other variables, the interactive transparency and illiquidity variables are insignificant throughout, with the exception of Model 3, where illiquidity enters at a 5% confidence level with an incorrect positive sign. The *S&P500* variable again also enters consistently with its expected negative sign at statistically significant levels. The biggest change is in the performance of the Google-based default proxy. This variable now enters with its predicted positive sign at statistically significant levels for all of our specifications. It seems that this proxy is more adept at picking up changes in default perception among the OECD country sub-sample.

This perception is confirmed for the non-OECD country sub-sample, which yields much weaker results (Table A5). In particular, the Google-based proxy enters with the incorrect, although usually insignificantly for the non-OECD sub-sample. This discrepancy with the OECD sub-sample may reflect the fact that this crisis hit wealthier countries harder than emerging market economies [e.g. Rose and Spiegel (2009a)]. It may also reflect the greater search volume found among OECD countries. Still, despite the poor performance of the default proxy, the remaining qualitative results are quite similar to those in the full sample.

4.3 Announcement Effects

We also examine the impact of the announcements listed in Table 1. We divide up the seven announcements listed into those applying to what we term the "major central banks," the ECB, the BOE, the SNB, the BOJ and the Bank of Canada, and those dealing with the central banks of other economies. We have three major bank announcement weeks: 1) The week including December 12, 2007, when the Federal Reserve initially announced the central bank swap programs with the ECB and the SNB, 2) the week including September 18, 2008, when swap lines were introduced with the BOJ, the BOE, the Bank of Canada and funds were increased for the ECB and the SNB, and 3) the week including October 13 and 14, 2008, when the ceilings on swap magnitudes were lifted

variable being 0.30

with the ECB, the BOE, the SNB, and the BOJ. We have two weeks with major announcements concerning other central banks, including September 24, 2008, when swap lines were introduced with Australia, Sweden, Denmark, and Norway, and the week of October 28 and 29, 2008, when swap lines were introduced with the reserve banks of New Zealand, Brazil, Mexico, Korea and Singapore.

Unlike the anticipated auctions examined above, we consider the "event week" associated with the announcements as the week in which the announcement was made. The intuition behind this assumption is that information flows are likely to be close to instantaneous, while the liquidity effects of anticipated injections of capital on other countries may take some time to establish.

We examine the impacts of these announcements by interacting our three variables of interest, *Exposure*, *Transp* and *Illiquid* with two announcement date dummies, labeled by the date of the first important announcement of that week. We also include the auction information included in the specifications above for completeness, but these coefficient values are suppressed for space reasons.³² The remainder is the same as our base specification.

The results for the major central bank announcements are shown in Table 7a. One can see that the impact of the announcements varied widely throughout the crisis. For the week of December 12, 2007, the interactive exposure variables are all insignificant. However, the transparency variables all now enter with their expected positive signs at statistically significant levels. Moreover, the interactive illiquidity variable enters with its expected negative sign throughout, although at statistically significant levels in only four of the seven specifications.

For the the week including October 13 and 14, 2008, when the ceilings on swaps with the major central banks were lifted, the interactive exposure variable enters negatively throughout, and is statistically significant for specifications, except Model 5. The interactive *Transp* variable again universally enters negatively at statistically significant levels. Moreover, the interactive illiquidity

³²These values are printed in Table A9.

variable usually obtains a negative sign, but is insignificant throughout. The similarity between these results and those of our base regressions is striking. Of course, this announcement also coincided with the height of the crisis, a time when TAF auction activity was also peaking. The similarities with the results for auction volumes and tenors is therefore not surprising.

We next turn to the announcements concerning swap arrangements with other central banks. These are shown in Table 7b. In the September 24, 2008 announcement, when swap lines were introduced with Australia, Sweden, Denmark, and Norway, we obtain negative coefficient estimates on the exposure variable throughout, but only at statistically significant levels in Model 5. However, we again obtain positive and statistically significant coefficient estimates on the interactive transparency variable throughout. The illiquidity measure is universally insignificant.

The final announcement, that of October 28 and 29, 2008, when swap lines were introduced with the reserve banks of New Zealand, Brazil, Mexico, Korea and Singapore, seems to be more similar to the October 13 announcement discuss above. The interactive exposure variables enter negatively throughout, and at 1% confidence levels for five of the seven specifications. The interactive transparency variable again enters negatively for all specifications at statistically significant levels throughout, while the illiquidity variable is mixed and insignificant for all specifications except Model 2.

Overall, the results for the announcement dates were mixed across event dates. The results for two of the announcement weeks – October 13 and 14, 2008, when the ceilings on swaps with the major central banks were lifted, and October 28 and 29, 2008, when swap lines were introduced with the reserve banks of New Zealand, Brazil, Mexico, Korea and Singapore – were very similar to those obtained for base specification of the actual auctions above. In particular, we obtained statistically significant coefficient estimates for all of our US exposure measures. However, for two of the other event weeks (that containing December 12, 2007 when the swap lines were originally introduced and that containing September 24, 2008, when the swap program was broadened to include Australia,

Sweden, Denmark, and Norway) the coefficient estimate on the interactive transparency variable entered for the first time with its predicted positive coefficient estimate at statistically significant levels.

It seems plausible that the results for the announcements in October were similar to those of the actual auctions because it was during that month that auction volume peaked. However, it seems difficult to draw parallels between the two event dates that yielded significant coefficient estimates for the interactive transparency variable for the first time. Both involved an expansion of the swap program, the first was the actual initiation of the program while the second expanded it beyond the major central banks. The significant coefficient estimate obtained for the transparency variable suggests that these expansions were of particular importance to countries with more opaque US asset holdings.

4.4 Differential Impacts for Swap Partner Countries

While the evidence above suggests that the broad cross section was affected by the international swap arrangements, it seems likely that the principal countries directly involved in those swaps may have been more affected on average. To investigate that possibility, we add slope dummies for countries that were direct auction recipients. We add a variable $direct_{t-1}$ that takes value the value of auction volume to country i times the average tenor of the securities auctioned at time $t - 1$ if country i received TAF funds in period $t - 1$, and 0 otherwise to capture the additional impact on CDS spreads of being a direct recipient of the TAF funds. We also add three interactive variables to our base specification: $Exposure_{it} \cdot direct_{t-1}$ which interacts the exposure variables with a variable $direct_{t-1}$, which takes the value of auction volume to country i times the average tenor of the securities auctioned at time $t - 1$ if country i received TAF funds in period $t - 1$, and 0 otherwise, $Transp_{it} \cdot direct_{t-1}$, and $Illiquidity_{it} \cdot direct_{t-1}$. These allow the direct effect to vary by country characteristics according to the predictions of the theory.

The results are shown in Table 8. The $direct_{t-1}$ variable obtains a positive sign throughout, but is insignificant at a 5% confidence level. The slope coefficients of the trade-related interactive direct exposure variables are negative throughout, except for Model 5, but are only statistically significant in Models 6 and 7. We also obtain negative, but usually insignificant coefficient estimates for the direct interactive illiquidity variables, with the exceptions again being Models 6 and 7, while the direct interactive transparency variable is insignificant throughout. The results for the overall variables from our base specification are little changed by the inclusion of these direct impact variables. In particular, the interactive exposure measures enter significantly with their predicted negative signs for all specifications except Model 5. As a result, we conclude that the international swaps did indeed serve to promote general dollar liquidity, and gave little measurable special assistance to those countries who were the direct recipients of the funds.

Finally, we next turn to the countries explicitly named in announcements concerning changes in the swap programs to examine if those countries exhibited additional sensitivity to country characteristics relative to the non-partner countries. We run our specification for announcements with the $Exposure_{it}$, $Transp_{it}$, and $Illiquidity_{it}$ variables interacted with two new variables $major$ and $other$. $major$ is a dummy variable that takes value one for the "major central banks," namely the ECB, the SNB, the BOE, the BOJ, and the Bank of Canada, on dates when they are specifically mentioned in Federal Reserve Announcements, and 0 otherwise, and take value 0 for the other central banks in our sample throughout. Similarly, dummy variable $other$ takes value one for the other central banks in our sample on dates when they are specifically mentioned in Federal Reserve Announcements, and value 0 otherwise, and value 0 for the major central banks throughout. We pool across these two groups of central banks because there are too few mentioned in any individual announcement to obtain an estimate of any extra sensitivity directly-named countries might have to these announcements. The cost of this aggregation is that we must constrain the coefficients to be identical across countries within these groups. We ran the specification including both the announcement events and the actual auction data, with the new interactive terms added. To

conserve on space, we only report the coefficient estimates on the slope coefficients, which can be found in Table 9.³³

The interactive exposure variables obtain negative coefficient estimates throughout, both for announcements involving major and other central banks. However, they are only statistically significant half of the time. The interactive transparency variable for swap announcements concerning major central banks is negative, and significant in four of the seven specifications. However, the coefficient for announcements concerning swap arrangements with other central banks, the interactive transparency variable is universally positive and statistically significant in six of the seven specifications. This suggests that among the non-major central banks countries, there was additional sensitivity to the opaqueness of US asset holdings among actual swap partners. Lastly, there was little observable difference in sensitivity to swap announcements by illiquidity among swap partner countries, as our coefficient estimates by this characteristic were mixed.

Overall, we did not observe much heterogeneity between the responsiveness of actual swap partners and the other countries in our sample, suggesting that the swaps acted more as a general injection of dollar liquidity worldwide than as funds that disproportionately assisted the countries towards whom these swaps were targeted. However, one notable exception was the interactive transparency variable for other central banks. Our previous results suggested that the September 24, 2008 announcement introducing swap lines to Australia, Denmark, Sweden and Norway disproportionately benefited countries with more opaque US portfolios. Our results in this section suggest that the swap partner countries were even more sensitive to asset opaqueness.

³³The full specification can be found in appendix Table A5.

5 Conclusion

This paper argues that the appreciation of the U.S. dollar exhibited at the height of turbulence during the recent global financial crisis suggests that there was a global dollar shortage. Models with illiquidity in dollar markets can mimic this behavior, as declines in some dollar asset values – as occurred to toxic US during the global financial crisis such as mortgage-backed securities – can result in the appreciation of other dollar assets that can serve as substitutes in the provision of liquidity services. This includes currency, which is a potential explanation of the surprising dollar exchange rate appreciation that occurred at the height of the global financial crisis. These models predict that injections of dollar liquidity, as occurred during the TAF auctions of the major foreign central banks, will have a disproportionately beneficial impact on economies that are more heavily exposed to the United States through trade or financial channels, have more opaque assets, or have deeper illiquidity problems.

We take these predictions to a cross-country panel, examining the impact of the TAF auctions on CDS spreads in a format that avoids a number of the problems encountered by the event studies in the existing literature. Our results suggest that the benefits of the TAF auctions were disproportionately enjoyed by those countries that had greater trade or asset exposure to the United States. We obtain weaker or incorrect results for national differences in the impact of the auctions by the transparency of dollar holdings and measures of illiquidity.

Looking at announcements concerning the TAF auctions, we found a discrepancy between those announcements that came at the height of the financial crisis and other announcements in our sample. For announcements in October 2008, we obtained results that were similar to those observed for the actual auctions throughout. In particular, we observed greater sensitivity to the announcements among countries that had greater trade or financial exposure to the United States. In contrast, for two of the other three announcements, we observed greater sensitivity among countries holding more opaque asset portfolios, again in keeping with the predictions of the theory.

Overall, our results suggest that the swap arrangements disproportionately benefited those countries that were more exposed to the United States, and we also obtain some evidence of disproportionate benefits to countries holding more opaque US asset portfolios. As suggested by theory, this is what one would expect from an effective dollar liquidity injection. Our results therefore support the claim that the swap arrangements provided tangible liquidity improvements. However, we should stress that we make no claims about the welfare implications of the swap arrangements here.

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6 Appendix

This appendix summarizes the model in Rose and Spiegel (2011), an international version of the search-based asset model of Lagos and Wright (2005), and its implications for the empirical specification in the paper. Detailed derivations of can be found in a technical appendix available online at <http://www.frbsf.org/economics/economists/mspiegel/wp11-18appendix.pdf>.

There are two countries in the model, u and r , which can be interpreted as representing the United States and the rest of the world with identical preferences and production technologies. Country z has world output share of τ_z ; ($z = u, r$), where $0 \leq \tau_z \leq 1$ and $\tau_u = 1 - \tau_r$.

In each period a continuum of infinitely lived agents participate in two distinct international markets: One is a Walrasian centralized global market, and another is a decentralized market, where pairs of buyers and sellers from the two countries are randomly matched. As in Lagos and Wright (2005) transactions in the decentralized market are characterized by a double-coincidence problem, which rules out barter, and anonymity, which rules out the provision of credit between matched agents. A tangible medium of exchange is therefore required for transactions to take place.

On each date, agents from country z ($z = u, r$) can produce a tradable homogeneous good for the centralized market, x , using labor, h_z , according to the production function $x_z = h_z$. Utility is assumed to be concave in x and negatively linear in h according to $U(x_z) - h_z$ and $U'(0) = \infty$, so that x_z^* , the optimal production of x in each country satisfies $U(x_z^*) = 1$.

Agents also produce a good, q_z , which is tradable in the international decentralized market. q_z is produced at disutility $c(q_z)$, where $c' > 0$, $c'' > 0$, and $c(0) = c'(0) = 0$. Agents value q_z according to the concave function $v(q_z)$, where $v' > 0$, $v'' < 0$, $v(0) = 0$, and $v'(0) = \infty$, so that q_z^* , the optimal production of q_z satisfies $v'(q_z) = c'(q_z)$. To highlight the role that differences in information sets and asset illiquidity play in determining outcomes, we assume that both x and q are homogeneous across countries.

Each country has a domestic money supply, which is in fixed supply, and a real asset that yields a dividend in the centralized market the following period. There are good and bad assets. Bad assets yield a zero dividend, while good assets yield a dividend of δ_z units of x ; $z = u, r$. Bad assets can be produced at zero cost.

Let ϕ_z and ψ_z represent the values of money and real assets of country z in the centralized market in terms of x respectively. As in Lester, Postlewaite, and Wright (2009b), agents can distinguish between bad and good assets in the centralized market, but in the decentralized market only informed agents can. Sellers who do not know the value of an asset will refuse to accept it at a positive price. As a result, bargaining only takes place where both agents are informed.

We focus on steady state equilibria. There is a fixed supply of trees in each country, A_z , and the supplies of both currencies grow at a constant rate, γ_z . Let \hat{k} represent the next period value of any variable k , so that $\hat{M}_z = \gamma_z M_z$. Agents worldwide are assumed to share a common discount factor, β , and we assume that $\gamma_z > 1 > \beta$ for both countries.³⁴

³⁴If allowed, agents may choose to keep some of their assets out of the bargaining process in the decentralized

All assets owned by agents are assumed to be brought into the decentralized market. We also assume that assets are "scarce," and therefore carry a liquidity value over their value in exchange the following day in the centralized market. The conditions needed for this assumption to hold are shown below.

Agents' centralized market portfolios are comprised of $m_{z,u}$ units of country u currency, $m_{z,r}$ units of country r currency, $a_{z,u}$ units of country u real assets, and $a_{z,r}$ units of country r assets. Let y_z represent income of an agent from country z in the centralized market, which satisfies

$$y_z = \phi_u m_{z,u} + \phi_r m_{z,r} + (\delta_u + \psi_r) a_{z,u} + (\delta_r + \psi_u) a_{z,r}. \quad (2)$$

Let $W(y_z)$ be the value function of an agent from country z in the centralized market, and define $V_z(m_{z,u}, m_{z,r}, a_{z,u}, a_{z,r})$ as the value function of an agent from country z in the decentralized market with portfolio $(m_{z,u}, m_{z,r}, a_{z,u}, a_{z,r})$. An agent from country z solves

$$\max_{x_z, h_z, \widehat{m}_{z,u}, \widehat{m}_{z,r}, \widehat{a}_{z,u}, \widehat{a}_{z,r}} W(y_z) = \{U(x_z) - h_z + \beta V_{z,u}(\widehat{m}_{z,u}, \widehat{m}_{z,r}, \widehat{a}_{z,u}, \widehat{a}_{z,r})\} \quad (3)$$

subject to

$$x_z \leq h_z + y_z - \phi_u \widehat{m}_{z,u} - \phi_r \widehat{m}_{z,r} - \psi_u(\widehat{a}_{z,u}) - \psi_r(\widehat{a}_{z,r}) + T_z, \quad (4)$$

where T_z is a lump-sum transfer returned to private agents in country z from revenues generated by money creation, $T_z = (\gamma_z - 1)M_z$. We demonstrate in Rose and Spiegel (2011) that y_z does not enter into the first order conditions. This is the mechanism through which the degenerate portfolio solutions are recovered each time the agents return to the centralized market in the Lagos and Wright (2005) framework. There are also four asset market clearing conditions, as the representative agent from each country holds his country's share of each asset.

In the decentralized market, agents are randomly paired into bilateral meetings. Let z and k represent the countries of origin of the buyer and seller respectively in the decentralized market ($z, k = u, r$). To allow for international liquidity differences we assume that sellers in the decentralized market only accept assets denominated in their domestic currencies. The probability of landing in a meeting in which there is a coincidence of wants is exogenous, but is proportional to the share of output of country k , τ_k , and is greater among agents originating from the same country by an exogenous parameter $\alpha > 1$. Let $\lambda_{z,k}$ represent the probability that an agent from z is paired with an agent from k from whom he wants to buy, and $\widetilde{\lambda}_{z,k}$ represent the chance that an agent from z is paired with an agent from country k to whom he wants to sell. We assume that $\lambda_{z,k} \equiv \lambda \tau_k$ when $z \neq k$ and $\lambda_{z,k} \equiv \lambda \alpha \tau_k$ when $z = k$, and that $\widetilde{\lambda}_{z,k} \equiv \widetilde{\lambda} \tau_k$ when $z \neq k$ and $\widetilde{\lambda}_{z,k} \equiv \widetilde{\lambda} \alpha \tau_k$ when $z = k$, where λ and $\widetilde{\lambda}$ are exogenous constants.

All agents from country k are informed about the value of their domestic currency, but only

market [Lagos and Rocheteau (2008)]. We rule this out for simplicity

a fraction ρ_k are informed about the value of a_k . We label meetings where the seller is informed about a_k "type 2," and meetings where the seller is uninformed "type 1." The type of meeting that is taking place is known to all.

Consider a type n meeting ($n = 1, 2$) where there is a coincidence of wants between a buyer from country z and a seller from country k . Let $p_{z,k,n}$ represent the price paid by the buyer from country z to a seller from country k for $q_{z,k,n}$ units of the good, and let y_z and y_k represent the wealth of the buyer and the seller respectively. Finally, let $\omega_{z,k,n}$ be the value of acceptable funds possessed by the buyer, i.e. those recognized by the seller and denominated in the seller's domestic currency. Given our assumptions above, $\omega_{z,k,1} = \phi_k m_{z,k}$, and $\omega_{z,k,2} = \phi_k m_{z,k} + (\psi_k + \delta_k) a_{z,k}$.

Assuming that the buyer has bargaining power θ and threat points are given by continuation values, the generalized Nash bargaining solution is similar to that in Lagos and Wright (2005):

$$\max_{q_{z,k,n}, p_{z,k,n}} [[v(q_{z,k,n}) + W(y_z - p_{z,k,n})] - W_z(y_z)]^\theta [[-c(q_{z,k,n}) + W(y_k + p_{z,k,n})] - W(y_k)]^{1-\theta} \quad (5)$$

subject to $p_{z,k,n} \leq \omega_{z,k,n}$.

The first order conditions satisfy

$$p_{z,k,n} = \frac{\theta v'(q_{z,k,n})c(q_{z,k,n}) + (1-\theta)v(q_{z,k,n})c'(q_{z,k,n})}{\theta v'(q_{z,k,n}) + (1-\theta)c'(q_{z,k,n})} \equiv \eta(q_{z,k,n}), \quad (6)$$

and

$$-\theta[-c(q_{z,k,n}) + p_{z,k,n}] + (1-\theta)[v(q_{z,k,n}) - p_{z,k,n}] - \varphi[-c(q_{z,k,n}) + p_{z,k,n}]^\theta [v(q_{z,k,n}) - p_{z,k,n}]^{(1-\theta)} = 0. \quad (7)$$

We assume that we are in the case where the liquidity constraint is binding, which implies that $p_{z,k,n} = \omega_{z,k,n}$ and $q_{z,k,n}$ satisfies 6 for $p_{z,k,n} = \omega_{z,k,n}$. Note that the terms of trade only depend on the buyer's portfolio.

The value function of an agent from country z in the decentralized market then satisfies

$$V_z = \sum_{n=1}^2 [\lambda_{u,n}[v(q_{z,u,n}) + W(y_z - p_{z,u,n})] + \lambda_{r,n}[v(q_{z,r,n}) + W(y_z - p_{z,r,n})]] + (1-\lambda)W(y_z) + \Psi_k \quad (8)$$

where $\lambda_{k,1} = \lambda_k(1 - \rho_k)$, $\lambda_{k,2} = \lambda_k\rho_k$, and Ψ_k represents the extra utility of an agent from country k associated with being a seller relative to having no trade opportunities.

Let $\tilde{q}_{z,k,n}$ and $\tilde{p}_{z,k,n}$ represent the volume and proceeds of selling q to an agent from country

z . Ψ_k satisfies

$$\Psi_k = \{\tilde{\lambda}_i[-c(\tilde{q}_{i,k,1}) + \tilde{p}_{i,k,1}] + \tilde{\lambda}_j[-c(\tilde{q}_{j,k,1}) + \tilde{p}_{j,k,1}]\}(1 - \Phi_k) + \{\tilde{\lambda}_i[-c(\tilde{q}_{i,k,2}) + \tilde{p}_{i,k,2}] + \tilde{\lambda}_j[-c(\tilde{q}_{j,k,2}) + \tilde{p}_{j,k,2}]\}\Phi_k \quad (9)$$

where Φ_k is an indicator variable that takes value 1 if agent k is informed about a_k , and 0 otherwise. It can be easily seen that Ψ_k is invariant to the portfolio decision of the agent from country k .

We follow Lagos and Wright (2005) in defining $\ell(q_{z,k,n}) \equiv [v'(q_{z,k,n})/\eta'(q_{z,k,n})] - 1$ as the liquidity premium in a type n meeting with a buyer from country z and a seller from country k . $\ell(q_{z,k,n})$ represents the increase in the buyer's utility from bringing an additional unit of country k currency into a type n meeting above its value in the next centralized market. We also follow Lagos and Wright (2005) in assuming that $\ell'(q_{z,k,n}) \leq 0$, which holds under usual conditions.

Differentiating V_z and combining the results with the centralized market solution conditions yields individual agents' demand for currencies u and r , as well as assets ψ_u and ψ_r . Equilibrium is defined as a solution for asset holdings, asset prices, the terms of trade in the decentralized markets, and leisure choices that satisfy the maximization conditions of each agent, the bargaining solutions in the decentralized markets, and market clearing in the centralized market. In the steady state, real variables are constant over time, and ϕ_z and M_z grow at a constant rate γ_z ($z = u, r$). The steady state versions of the money and asset demand equations satisfy

$$\frac{\gamma - \beta}{\beta\lambda_u} \geq (1 - \rho_u)\ell(q_{z,u,1}) + \rho_u\ell(q_{u,2}), \quad (10)$$

$$\frac{\gamma - \beta}{\beta\lambda_r} \geq (1 - \rho_r)\ell(q_{z,r,1}) + \rho_r\ell(q_{r,2}), \quad (11)$$

$$\frac{(1 - \beta)\psi_u - \beta\delta_u}{\beta(\psi_u + \delta_u)\lambda_u} = \rho_u\ell(q_{z,u,2}), \quad (12)$$

and

$$\frac{(1 - \beta)\psi_r - \beta\delta_r}{\beta(\psi_r + \delta_r)\lambda_r} = \rho_r\ell(q_{z,r,2}), \quad (13)$$

where the conditions hold with equality when positive levels of money or assets are held. Existence and uniqueness of the equilibrium is demonstrated in Rose and Spiegel (2011).

We next examine the comparative static impact of a decline in δ_u . First by equation 12, the change in ψ_u with a decline in δ_u satisfies

$$\frac{\partial\psi_u}{\partial\delta_u} = \frac{\delta_u - \beta(\psi_u + \delta_u)\lambda_u\rho_u\ell'(q_{z,u,2})I\{\omega_{z,u,2} < \eta(q^*)\}}{\psi_u - \beta(\psi_u + \delta_u)\lambda_u\rho_u\ell'(q_{z,u,2})I\{\omega_{z,u,2} < \eta(q^*)\}}. \quad (14)$$

The numerator of equation 14 is unambiguously positive, but the denominator is ambiguous in sign. The necessary condition for $\partial\psi_u/\partial\delta_u \geq 0$ is that $\ell'(q_{z,u,2})$ is not "too large". We require the restriction $\psi_u \geq \beta(\psi_u + \delta_u)\lambda_u\rho_u\ell'(q_{z,u,2})I\{\omega_{z,u,2} < \eta(q^*)\}$, which implies that asset values fall with declines in their dividend streams, which we adopt.

Substituting from equation 12 into equation 10 and totally differentiating with respect to $\phi_um_{z,u}$ and δ_u yields

$$\frac{\partial\phi_um_{z,u}}{\partial\delta_u} = \frac{\psi_u + \delta_u \frac{\partial\psi_u}{\partial\delta_u}}{(\psi_u + \delta_u)^2\beta\lambda_u(1 - \rho_u)\ell'(q_{z,u,1})I\{\widehat{\omega}_{z,u,1} < \eta(q^*)\}} \leq 0, \quad (15)$$

as $\frac{\partial\psi_u}{\partial\delta_u}$ can be signed as positive given the restriction above.

In contrast, it can be seen by inspection of equations 13 and 11 that ψ_r and $\phi_rm_{z,r}$ are invariant to changes in δ_u . This implies that a decline in δ_u will lead to an appreciation in country u 's exchange rate, ϕ_u/ϕ_r .

A formal proof is in Rose and Spiegel (2011). Intuitively, as $m_{z,u}$ is exogenous, any change in real balances, $\phi_um_{z,u}$ must come from an increase in ϕ_u . In contrast, since $\phi_rm_{z,r}$ is unchanged ϕ_r must be unchanged as well.

We next turn to our empirical work by assessing the implications of our model for the predicted impact of the U.S. dollar auctions by foreign central banks. We assume that dollar injections are surprise interventions that take place in the decentralized market subsequent to the fall in δ_u . As was the case empirically, the dollar injections are assumed to be loans which for tractability we assume are repaid in the next entry into the centralized market.

The impact of the liquidity injection on an agent from a foreign country can then be represented in terms of the change in the decentralized market value function with an increase in US currency holdings. The influence of country characteristics on the relative impact of the injections can then be shown in terms of the cross-partials of the decentralized market value function with respect to dollar holdings and these characteristics. We examine three: $\lambda_{r,u}$, ρ_u , and $\ell(q_{r,u,1})$. By equation 8, $\partial^2V_r/\partial m_{r,u}\partial\lambda_{r,u}$, $\partial^2V_r/\partial m_{r,u}\partial\rho_u$ and $\partial^2V_r/\partial m_{r,u}\partial\ell(q_{r,u,1})$ satisfy

$$\frac{\partial^2V_r}{\partial m_{r,u}\partial\lambda_{r,u}} = \phi_u[(1 - \rho_u)\ell(q_{r,u,1})I_{r,u,1} + \rho_u\ell(q_{r,u,2})I_{r,u,2}] \geq 0, \quad (16)$$

$$\frac{\partial^2V_r}{\partial m_{r,u}\partial\rho_u} = \phi_u\lambda_{r,u}[-\ell(q_{r,u,1})I_{r,u,1} + \ell(q_{r,u,2})I_{r,u,2}] < 0, \quad (17)$$

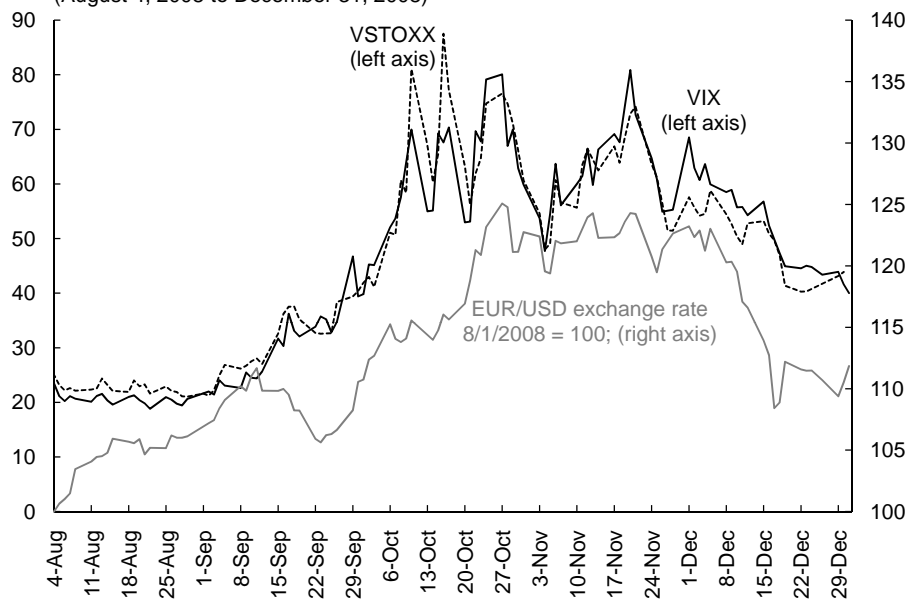
and

$$\frac{\partial^2V_r}{\partial m_{r,u}\partial\ell(q_{r,u,1})} = \theta_u\lambda_{r,u,1}I_{r,u,1} > 0. \quad (18)$$

Our interpretation is of these results is discussed in the text.

Figure 1: Stock Market Volatility and Bilateral Exchange Rate

American and European VIX indices; Euro-US dollar exchange rate index
(August 4, 2008 to December 31, 2008)

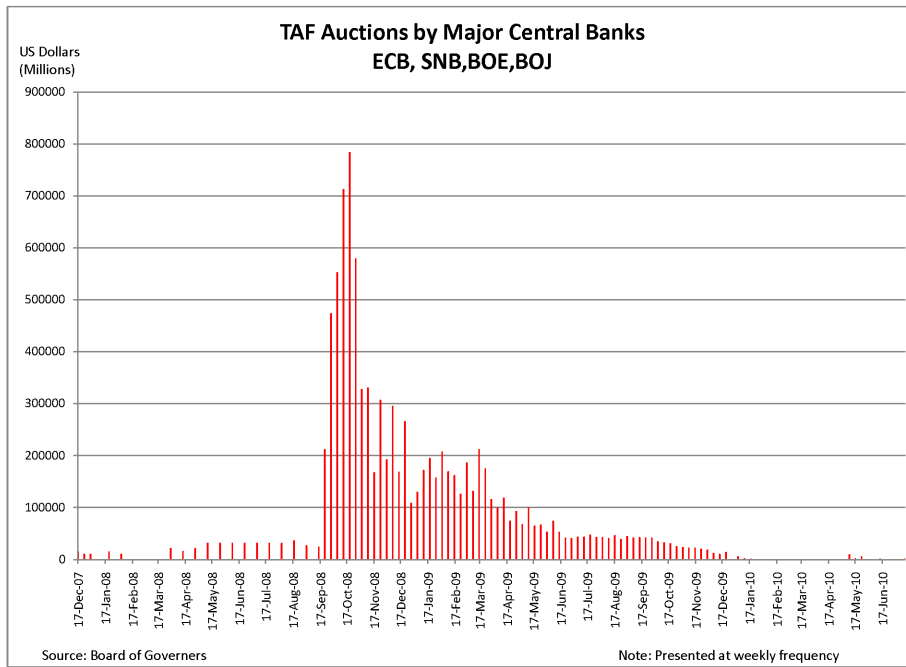


Sources: Bloomberg and Federal Reserve DistFAME

VIX and VSTOXX indices of equity market volatility in United States and European exchanges respectively. Daily frequency.

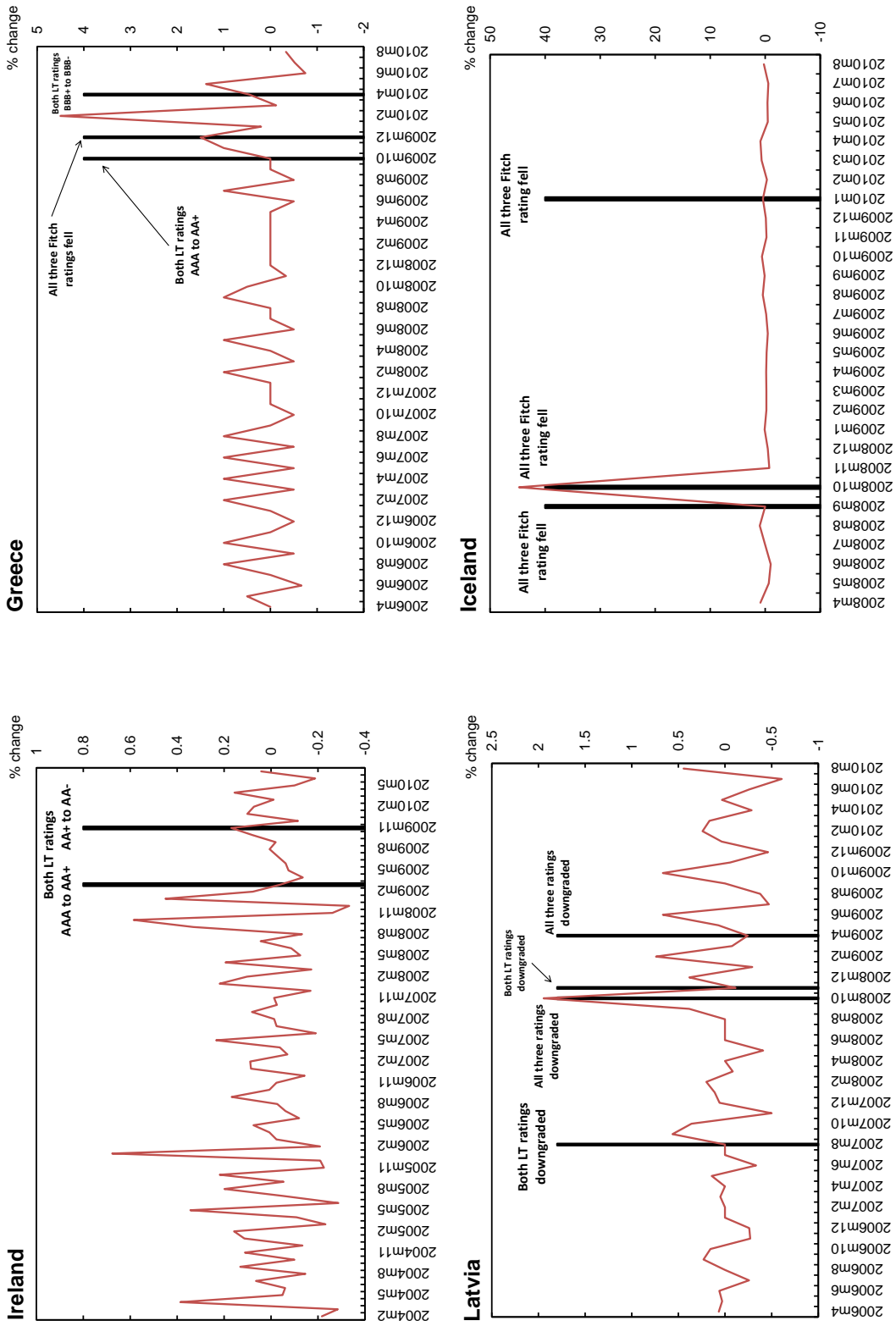
Dollar-euro exchange rate indexed to 100 on August 1, 2008.

Figure 2: Weekly Major Central Bank TAF Auction Volumes



Source: Board of Governors
Total weekly volume of TAF auctions by major central banks.

Figure 3: Change in Google index and changes in Fitch ratings



Plot of search volume in Google for identified countries and one or more of terms crisis, freeze, financial and recession. Source: Google and Fitch ratings

Table 1: Announcements concerning international swap arrangements: 2007-2008

Announcement Type	Date	Details
Swap lines introduced with ECB and SNB	12/12/2007	Authorization of temporary reciprocal currency arrangements with the European Central Bank and Swiss National Bank, providing \$20 billion and \$4 billion respectively. Arrangements authorized for up to six months.
Swap lines introduced with BOJ, BOE, and Bank of Canada. Funds increased for the ECB and SNB.	9/18/2008	New swap facilities authorized with the Bank of Japan for amounts up to \$60 billion, the Bank of England for up to \$40 billion, and the Bank of Canada for up to \$10 billion. In addition, swap line provisions increased to \$110 for the European Central Bank and \$27 billion to the Swiss National Bank. Arrangements authorized through January 30, 2009
Swap lines introduced with Australia, Sweden, Denmark, and Norway.	9/24/2008	Currency swap facilities established in amounts of up to \$10 billion each by the Reserve Bank of Australia and Sveriges Riksbank and amounts of up to \$5 billion each by Danmarks Nationalbank and the Norges Bank. Arrangements authorized through January 30, 2009
Unlimited swaps announced with ECB, BOE, and SNB.	10/13/2008	Funds provided to swap facilities with the Bank of England, the European Central Bank, and the Swiss National Bank increased to "accommodate whatever quantity of U.S. dollar funding is demanded". Arrangements were authorized through April 30, 2009.
Unlimited Swaps announced with the BOJ	10/14/2008	Unlimited swap arrangement announced with the Bank of Japan a day after the same announcement was made regarding swap lines with major central bank in Europe. Arrangements authorized through April 30, 2009
Swap lines introduced with the Reserve Bank of New Zealand	10/28/2008	New swap line of up to \$15 billion with the Reserve Bank of New Zealand announced. Arrangement authorized through April 30, 2009.
Swap Line Introduced with Brazil, Mexico, Korea, and Singapore.	10/29/2008	New swap facilities established in amounts of up to \$30 billion each by the Banco Central do Brazil, the Banco de Mexico, the Bank of Korea, and the Monetary Authority of Singapore. Authorized through April 30, 2009.

Source: Federal Reserve Board of Governors

Table 2: TAF Auctions by major central banks

Quarter	ECB		BOE		BOJ		SNB	
	Dollars Auctioned (Millions)	Average Tenor (Days)	Dollars Auctioned (Millions)	Average Tenor (Days)	Dollars Auctioned (Millions)	Average Tenor (Days)	Dollars Auctioned (Millions)	Average Tenor (Days)
2007q4	20000	31.5	-	-	-	-	4000	28.0
2008q1	35000	28.0	-	-	-	-	10000	28.0
2008q2	130000	28.0	-	-	-	-	30000	28.0
2008q3	589742	11.3	216044	2.1	29622	28.0	132139	11.8
2008q4	3608841	8.3	667737	9.7	205635	54.2	196948	11.3
2009q1	1937722	7.4	30956	50.6	106253	59.1	51702	20.0
2009q2	865642	9.5	3503	59.7	36243	49.4	41006	18.7
2009q3	542729	7.5	538	26.4	8100	31.5	18	28.0
2009q4	259478	7.5	52	7.0	1300	32.4	-	-
2010q1	6575	7.0	-	-	100	29.0	-	-

Source: Federal Reserve Board of Governors

Notes: Size (millions of dollars) and average tenor (days) of dollar auctions by major central banks.

Table 3: Google searches and sovereign ratings

I. Full Sample: January 2004-September 2010

Rating	Coefficient	T-stat	# of Obs	# of Countries	R-squared
Local LT	-0.04**	-4.0	5360	100	0.07
Foreign LT	-0.01**	-5.15	5410	99	0.04
Foreign ST	-0.05**	-4.73	5363	98	0.09

II. Study Sample: December 2007-December 2009

Rating	Coefficient	T-stat	# of Obs	# of Countries	R-squared
Local LT	-0.04**	-9.7	2245	96	0.15
Foreign LT	-0.02**	-11.56	2270	96	0.08
Foreign ST	-0.06**	-10.81	2260	96	0.22

Regression of Fitch Ratings changes on changes in Google Index

Source: Google Insight for Search, Fitch

OLS panel estimation with country and time fixed effects included.

Monthly data with robust standard errors.

Table 4: Base specification

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-2.87*	-2.56**	-3.14**	-0.64**	-0.01	-0.66**	-0.68**
	(-2.15)	(-4.94)	(-2.95)	(-3.99)	(-1.42)	(-3.50)	(-3.62)
Transp*auction	-0.58*	-0.61*	-0.54*	-0.58*	-0.54*	-0.74**	-0.75**
	(-2.10)	(-2.65)	(-2.14)	(-2.43)	(-2.19)	(-3.26)	(-3.27)
Illiquid*auction	1.64	0.68	0.81	0.94	0.51	0.89	0.94
	(1.04)	(1.25)	(1.22)	(1.80)	(0.66)	(1.46)	(1.53)
Exposure*SP500	-6.97**	-6.05**	-8.16**	-1.85**	-0.03**	-2.00**	-1.98**
	(-5.20)	(-4.58)	(-6.20)	(-21.05)	(-4.87)	(-15.37)	(-14.61)
Δ Default	0.36	0.48	0.28	0.61	1.51*	0.66	0.67
	(0.52)	(0.76)	(0.43)	(0.85)	(2.11)	(0.91)	(0.93)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.200	0.200	0.212	0.240	0.209	0.217	0.216

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table 5: Alternative illiquidity measures

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
<u>I. Illiquidity proxied by ST Debt/GDP</u>							
Exposure*auction	-2.66*	-2.57**	-3.19**	-0.63**	-0.01	-0.64**	-0.66**
	(-2.42)	(-4.90)	(-2.88)	(-3.96)	(-1.45)	(-3.55)	(-3.66)
Transp*auction	-0.57*	-0.59*	-0.52*	-0.56*	-0.53*	-0.72**	-0.72**
	(-2.08)	(-2.62)	(-2.06)	(-2.39)	(-2.14)	(-3.24)	(-3.25)
ill_gdp*auction	0.59	0.32	0.56	0.35	0.32	0.26	0.23
	(1.16)	(1.20)	(1.46)	(1.14)	(0.91)	(1.00)	(0.88)
<u>II. Illiquidity proxied by ST Debt/Reserves</u>							
Exposure*auction	-2.60*	-2.55**	-3.14**	-0.62**	-0.01	-0.64**	-0.66**
	(-2.51)	(-4.94)	(-2.92)	(-3.91)	(-1.46)	(-3.52)	(-3.63)
Transp*auction	-0.58*	-0.59*	-0.53*	-0.57*	-0.54*	-0.73**	-0.73**
	(-2.12)	(-2.66)	(-2.09)	(-2.42)	(-2.18)	(-3.27)	(-3.28)
ill_res*auction	0.09	0.02	0.09	0.00	0.04	-0.00	-0.01
	(0.85)	(0.22)	(0.95)	(0.02)	(0.44)	(-0.02)	(-0.12)
<u>III. Illiquidity proxied by Greenspan-Guidotti measure</u>							
Exposure*auction	-2.18*	-2.09**	-2.49*	-0.44*	-0.01	-0.35	-0.39
	(-2.13)	(-4.04)	(-2.23)	(-2.16)	(-1.03)	(-1.19)	(-1.40)
Transp*auction	-0.18	-0.24	-0.20	-0.26	-0.18	-0.20	-0.24
	(-0.57)	(-0.74)	(-0.65)	(-0.72)	(-0.51)	(-0.43)	(-0.54)
ill_GG*auction	0.32	0.30	0.29	0.28	0.31	0.40	0.37
	(1.75)	(1.51)	(1.51)	(1.14)	(1.46)	(1.27)	(1.23)

Coefficients for ill_gdp are presented in millions

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table 6: OECD countries only

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-6.53*	-2.42**	-4.87**	-0.96**	-0.01*	-1.00**	-0.93**
	(-2.76)	(-3.19)	(-3.44)	(-3.63)	(-2.08)	(-4.09)	(-3.83)
Transp*auction	-0.01	-0.22	-0.05	-0.12	-0.07	-0.31	-0.33
	(-0.04)	(-0.75)	(-0.16)	(-0.48)	(-0.33)	(-1.12)	(-1.18)
Illiquid*auction	5.20	1.14	3.45*	1.45	1.55	0.76	0.50
	(2.02)	(1.05)	(2.43)	(1.60)	(1.15)	(0.77)	(0.49)
Exposure*SP500	-10.98**	-4.79**	-8.75**	-2.05**	-0.04**	-2.32**	-2.08**
	(-5.07)	(-3.72)	(-5.78)	(-8.34)	(-5.58)	(-5.37)	(-4.91)
Δ Default	2.60**	2.56**	2.53**	2.93**	2.66**	2.79**	2.80**
	(4.00)	(4.15)	(4.04)	(5.06)	(4.31)	(4.58)	(4.59)
Num of Obs	2220	2220	2220	2220	2220	2220	2220
R^2	0.188	0.184	0.195	0.197	0.200	0.186	0.182

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table 7a: Swap announcements with major central banks

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
<u>I. Dec 12 '07: Lines introduced with ECB and SNB</u>							
Exposure	-14.76 (-1.62)	5.18 (0.66)	-3.48 (-0.37)	-2.69 (-0.87)	0.01 (0.25)	-0.56 (-0.19)	-0.91 (-0.30)
Transp	18.21** (3.33)	14.85** (2.85)	15.46** (2.89)	17.52** (3.30)	14.30* (2.56)	17.83** (3.47)	18.14** (3.49)
Illiquid	-33.20 (-1.52)	-22.12* (-2.02)	-23.45 (-1.91)	-27.81 (-1.49)	-30.93** (-3.05)	-41.52* (-2.60)	-39.76* (-2.39)
<u>II. Sep 18 '08: Lines introduced with BOJ, BOE, and Bank of Canada</u>							
Exposure	-3.40 (-0.10)	-25.09 (-1.36)	-18.37 (-0.61)	-8.66 (-1.74)	-0.12 (-1.06)	-7.65 (-1.29)	-8.33 (-1.41)
Transp	-14.05* (-2.34)	-11.44* (-2.12)	-11.02 (-1.92)	-6.54 (-0.99)	-9.07 (-1.40)	-10.62 (-1.78)	-10.72 (-1.81)
Illiquid	-50.36 (-0.91)	-55.20 (-1.76)	-51.05 (-1.62)	-24.06 (-0.59)	-35.95 (-0.94)	-19.98 (-0.43)	-18.84 (-0.39)
<u>III. Oct 13 '08: Unlimited swaps with ECB, BOE, BOJ, and SNB</u>							
Exposure	-85.29** (-4.50)	-80.45** (-5.04)	-102.39** (-5.56)	-14.05* (-2.22)	-0.05 (-0.13)	-17.45* (-2.51)	-17.99* (-2.64)
Transp	-25.29** (-3.15)	-24.10** (-3.13)	-21.22** (-2.70)	-31.06** (-3.84)	-30.52* (-2.57)	-37.98** (-4.89)	-38.17** (-4.92)
Illiquid	65.14 (1.76)	-51.83 (-1.65)	-25.54 (-0.60)	-36.31 (-1.30)	-71.59 (-1.08)	-16.50 (-0.47)	-16.55 (-0.46)
Num of Obs	4000	4000	4000	4000	3601	4000	4000
R^2	0.309	0.313	0.319	0.356	0.318	0.343	0.343

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

Major central banks include: European Central Bank, Swiss National Bank, Bank of Japan, Bank of England, and Bank of Canada.

* $p < 0.05$, ** $p < 0.01$

Table 7b: Swap announcements with other central banks

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
I. Sep 24 '08:							
Lines introduced with Australia, Sweden, Denmark, and Norway							
Exposure	-42.89 (-1.24)	-39.94 (-1.15)	-42.76 (-1.26)	-7.40 (-0.75)	-0.39* (-2.07)	5.79 (0.58)	4.94 (0.49)
Transp	97.17** (4.99)	96.36** (5.02)	97.26** (4.97)	96.73** (4.83)	108.09** (5.21)	90.89** (5.09)	91.06** (5.11)
Illiquid	88.06 (0.26)	61.04 (0.18)	64.72 (0.19)	64.69 (0.19)	106.20 (0.31)	5.76 (0.02)	10.83 (0.03)
II. Oct 28 '08:							
Lines introduced with New Zealand, Brazil, Mexico, South Korea, and Singapore							
Exposure	-73.09 (-1.62)	-74.99** (-3.54)	-94.30** (-2.85)	-44.46** (-8.11)	-0.29 (-1.10)	-46.21** (-6.79)	-46.03** (-6.69)
Transp	-71.79** (-6.44)	-69.06** (-6.18)	-63.61** (-5.94)	-50.05** (-7.43)	-60.96** (-4.23)	-75.93** (-8.54)	-76.98** (-8.72)
Illiquid	67.53 (0.52)	-133.56* (-2.09)	-85.45 (-1.24)	34.15 (0.52)	-75.38 (-0.86)	81.06 (0.85)	71.73 (0.69)
Num of Obs	4000	4000	4000	4000	3601	4000	4000
R^2	0.309	0.313	0.319	0.356	0.318	0.343	0.343

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

Other central banks include: Reserve Bank of Australia, Danmarks Nationalbank, Bank of Korea, Bank of Mexico, Norges Bank, Sveriges Riksbank.

* $p < 0.05$, ** $p < 0.01$

Table 8: Base with direct auction

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
direct	1.78 (1.81)	1.81 (1.90)	1.77 (1.82)	1.49 (1.49)	1.91 (2.01)	1.67 (1.67)	1.76 (1.77)
Exposure*auction	-2.82* (-2.08)	-2.62** (-5.22)	-3.15** (-2.92)	-0.68** (-3.86)	-0.01 (-1.54)	-0.63** (-3.17)	-0.65** (-3.27)
Exposure*dir	-4.70 (-0.24)	-17.02 (-1.84)	-16.26 (-0.97)	-1.73 (-1.34)	0.01 (0.10)	-1.46** (-2.74)	-1.41** (-2.84)
Transp*auction	-0.39 (-1.23)	-0.42 (-1.50)	-0.35 (-1.17)	-0.34 (-1.03)	-0.32 (-0.97)	-0.58* (-2.07)	-0.58* (-2.08)
Transp*dir	-0.33 (-0.21)	0.08 (0.09)	0.19 (0.14)	-1.00 (-1.24)	-0.74 (-0.65)	-0.70 (-1.00)	-0.72 (-1.03)
Illiquid*auction	1.26 (0.79)	0.33 (0.50)	0.44 (0.59)	0.56 (0.93)	0.33 (0.40)	0.48 (0.73)	0.53 (0.80)
Illiquid*dir	-254.49 (-1.72)	-103.10 (-1.09)	-132.89 (-0.95)	-150.13 (-1.61)	-282.19 (-1.59)	-198.59** (-2.71)	-200.98** (-2.75)
Exposure*SP500	-6.97** (-5.20)	-6.03** (-4.57)	-8.16** (-6.19)	-1.85** (-21.01)	-0.03** (-4.86)	-1.99** (-15.35)	-1.98** (-14.60)
Δ Default	0.35 (0.51)	0.48 (0.76)	0.28 (0.43)	0.61 (0.85)	1.50* (2.08)	0.66 (0.90)	0.67 (0.92)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.201	0.201	0.213	0.241	0.210	0.218	0.217

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table 9: Slope coefficients for recipients in announcements

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*major	-0.23 (-0.43)	-3.93** (-4.91)	-1.02 (-0.97)	-0.61** (-3.27)	-0.01 (-1.99)	-0.50** (-5.73)	-0.49** (-5.55)
Transp*major	-0.09 (-0.68)	-0.23** (-2.99)	-0.06 (-0.49)	-0.39* (-2.27)	-0.05 (-0.40)	-0.47** (-3.28)	-0.47** (-3.17)
Illiquid*major	-21.87** (-3.57)	-2.93 (-0.29)	-17.83* (-2.57)	2.74 (0.27)	0.44 (0.03)	-21.73** (-5.48)	-22.10** (-5.57)
Exposure*other	-6.89* (-2.20)	-0.42 (-0.79)	-1.67 (-1.87)	-1.16** (-6.95)	0.02 (1.21)	2.39** (3.88)	0.05 (0.03)
Transp*other	2.01** (2.93)	2.00** (15.65)	2.69** (20.61)	1.29** (3.86)	3.00** (7.63)	5.44** (6.53)	2.91 (1.61)
Illiquid*other	-34.07 (-0.93)	146.83* (2.08)	134.03* (2.56)	49.01* (2.57)	-25.10 (-0.35)	-135.02** (-3.42)	58.81 (0.51)
Num of Obs	4000	4000	4000	4000	3601	4000	4000
R^2	0.232	0.230	0.242	0.269	0.240	0.246	0.244

Coefficients in table are presented in hundreds

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

Major central banks include:

European Central Bank, Swiss National Bank, Bank of Japan, Bank of England, and Bank of Canada.

* $p < 0.05$, ** $p < 0.01$

Table A1: Add Level Effects to Interaction Terms

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-2.52*	-2.49**	-3.00**	-0.63**	-0.01	-0.68**	-0.69**
	(-2.01)	(-4.90)	(-2.95)	(-3.98)	(-1.35)	(-3.70)	(-3.86)
TranspXauction	-0.61*	-0.61*	-0.55*	-0.59*	-0.56*	-0.74**	-0.74**
	(-2.25)	(-2.64)	(-2.19)	(-2.52)	(-2.27)	(-3.20)	(-3.25)
IlliquidXauction	0.63	-0.13	0.13	0.31	-0.27	0.35	0.41
	(0.37)	(-0.19)	(0.16)	(0.51)	(-0.28)	(0.48)	(0.55)
Exposure	-4.52	-6.91	-7.60	10.44	0.05	2.55*	2.24*
	(-1.46)	(-0.87)	(-1.01)	(1.98)	(0.98)	(2.50)	(2.13)
Transp	1.78	2.49	1.82	2.60	4.90	5.01	3.94
	(0.66)	(0.88)	(0.67)	(0.79)	(1.78)	(1.33)	(1.23)
Illiquid	14.24*	15.24*	12.63*	13.65	15.14*	11.42	11.17
	(2.10)	(2.61)	(2.26)	(1.53)	(2.04)	(1.75)	(1.61)
ExposureXSP500	-6.99**	-6.05**	-8.17**	-1.85**	-0.03**	-2.00**	-1.98**
	(-5.21)	(-4.58)	(-6.19)	(-20.91)	(-4.88)	(-15.31)	(-14.59)
Chg_Default	0.37	0.48	0.29	0.61	1.52*	0.66	0.67
	(0.53)	(0.76)	(0.44)	(0.85)	(2.12)	(0.91)	(0.92)
<i>N</i>	4000	4000	4000	4000	3601	4000	4000
<i>R</i> ²	0.201	0.200	0.212	0.241	0.209	0.217	0.216

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A2: Base specification without time dummies

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
auction	0.38 (1.69)	0.49* (2.26)	0.43 (1.72)	0.25 (0.91)	0.32 (1.21)	-0.04 (-0.10)	0.05 (0.14)
Exposure*auction	-1.54 (-1.30)	-2.49** (-3.89)	-2.34 (-1.68)	-0.29 (-1.52)	-0.00 (-0.88)	-0.01 (-0.02)	-0.10 (-0.27)
Transp*auction	-0.26 (-0.63)	-0.33 (-0.83)	-0.30 (-0.72)	-0.14 (-0.32)	-0.27 (-0.57)	0.26 (0.43)	0.15 (0.27)
Illiquid*auction	-1.53 (-1.03)	-0.29 (-0.39)	-0.49 (-0.62)	-0.80 (-0.79)	-0.76 (-0.80)	-1.52 (-0.91)	-1.32 (-0.85)
Exposure*SP500	-8.82** (-5.75)	-7.59** (-4.62)	-10.05** (-6.72)	-2.17** (-22.59)	-0.04** (-5.54)	-2.44** (-16.93)	-2.43** (-16.09)
Δ Default	0.65 (0.68)	0.79 (0.88)	0.54 (0.60)	0.90 (0.94)	2.19* (2.43)	1.01 (1.01)	1.02 (1.02)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.108	0.106	0.126	0.164	0.122	0.130	0.129

t statistic in parentheses; OLS with country fixed effects and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A3: Single Word Google Search Results, Foreign Long-Term Debt

Word	Coeff	T-stat	R-sq	# of countries	# of obs
arrears	0.00263	0.63969	0.00007	4	89
balance	-0.00804	-0.78951	0.00037	73	3103
bankruptcy	0.00482	0.34133	0.00010	48	1081
bond	-0.03614	-1.37083	0.00767	83	3342
collapse	-0.00289	-1.01006	0.00030	56	1140
contraction	-	-	-	5	77
credit	-0.04056	-1.14338	0.00494	100	4705
crisis	-0.01548**	-30.78583	0.07466	96	3779
currency	-0.01646	-1.11072	0.00107	105	6703
danger	-0.00613	-1.33332	0.00079	52	1352
debt	-0.01316	-1.15926	0.00142	82	2866
deficit	0.01410	1.15733	0.00159	56	1443
depression	0.00430	0.38199	0.00012	55	1814
disaster	-0.00028	-0.35360	0.00000	68	2321
downturn	0.00212	0.88936	0.00002	10	159
equity	-0.00429	-1.03095	0.00015	67	2694
exposure	-0.01607	-0.71277	0.00163	34	769
failure	-0.01601	-0.53411	0.00175	62	2014
financial	-0.06121*	-2.26033	0.03905	100	4910
freeze	-0.02815**	-3.51057	0.00332	43	807
insolvency	0.00227	0.89918	0.00004	15	356
liability	-0.00110	-0.07959	0.00000	41	1136
liquidity	0.01442	1.23534	0.00092	16	332
market	-0.03413	-1.07742	0.00426	104	5908
overdue	-	-	-	2	33
recession	0.00103	0.28506	0.00002	55	910
risk	-0.01965	-1.01747	0.00172	86	3503
safety	-0.01318	-1.24273	0.00058	93	4365
shortage	-0.00072	-0.27161	0.00002	42	1005
stock	-0.04151	-1.52674	0.02512	97	5093
tax	-0.01440	-1.00984	0.00067	101	5635
unpaid	-0.01309	-1.27611	0.00037	13	230

OLS estimation with robust standard errors.

Table A4: Base specification with weekly time dummies

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-0.34 (-0.28)	-0.84 (-1.67)	-0.79 (-0.65)	-0.05 (-0.25)	0.00 (0.30)	0.26 (0.77)	0.15 (0.48)
Transp*auction	-0.13 (-0.36)	-0.16 (-0.46)	-0.16 (-0.45)	-0.09 (-0.22)	-0.11 (-0.27)	0.27 (0.49)	0.15 (0.30)
Illiquid*auction	0.77 (0.52)	0.30 (0.44)	0.41 (0.53)	0.22 (0.30)	0.27 (0.32)	-0.33 (-0.26)	-0.14 (-0.13)
Exposure*SP500	0.22 (0.24)	-0.47 (-0.74)	-0.08 (-0.07)	-0.44** (-3.05)	0.00 (0.54)	-0.37* (-2.22)	-0.41* (-2.56)
Chg_Default	-0.08 (-0.19)	-0.11 (-0.25)	-0.09 (-0.23)	-0.09 (-0.20)	0.49 (1.00)	-0.08 (-0.19)	-0.08 (-0.19)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.522	0.522	0.522	0.524	0.534	0.523	0.524

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, weekly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A5: Google default proxy generated from ad hoc word list

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-2.84*	-2.62**	-3.20**	-0.66**	-0.01	-0.70**	-0.72**
	(-2.13)	(-4.82)	(-3.00)	(-4.29)	(-1.50)	(-3.93)	(-4.03)
TranspXauction	-0.55*	-0.57*	-0.51*	-0.54*	-0.50*	-0.72**	-0.72**
	(-2.02)	(-2.51)	(-2.02)	(-2.29)	(-2.02)	(-3.18)	(-3.19)
IlliquidXauction	1.57	0.66	0.80	1.00	0.51	1.00	1.05
	(0.99)	(1.22)	(1.19)	(1.94)	(0.65)	(1.70)	(1.77)
ExposureXSP500	-6.97**	-6.04**	-8.13**	-1.79**	-0.03**	-1.94**	-1.92**
	(-5.22)	(-4.63)	(-6.26)	(-17.85)	(-5.03)	(-14.40)	(-13.63)
chg_google_old	0.01	0.01	0.01	0.01	0.02**	0.01	0.01
	(1.05)	(1.38)	(1.00)	(1.29)	(3.17)	(1.36)	(1.37)
N	4223	4223	4223	4223	3813	4223	4223
R^2	0.202	0.201	0.213	0.239	0.210	0.217	0.216

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A6: Exposure as share of GDP

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-3.79 (-2.00)	-10.76** (-5.45)	-4.13* (-2.39)	-0.87* (-2.48)	-0.26 (-1.42)	-0.44 (-1.44)	-0.75 (-1.47)
TranspXauction	-0.68** (-2.82)	-0.44 (-1.88)	-0.50* (-2.03)	-0.77** (-3.38)	-0.72** (-3.22)	-0.73** (-3.22)	-0.70** (-3.07)
IlliquidXauction	0.19 (0.39)	3.35* (2.22)	1.15 (1.26)	2.06 (1.77)	0.28 (0.41)	0.53 (0.79)	0.64 (0.89)
ExposureXSP500	-17.25** (-3.61)	-19.61** (-4.17)	-12.19** (-5.50)	-2.00** (-3.18)	-1.43* (-2.39)	-1.75* (-2.40)	-2.89* (-2.35)
Chg_Default	0.52 (0.72)	0.70 (1.02)	0.52 (0.72)	0.86 (1.28)	1.83** (3.04)	1.84** (3.04)	1.84** (3.04)
<i>N</i>	3953	3953	3953	3953	3555	3542	3465
<i>R</i> ²	0.185	0.202	0.210	0.173	0.162	0.160	0.163

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A7: BIS Bank Exposure Measure

	(1)	(2)
	US Bank Exp	BIS_GDP
ExposureXauction	-2.76** (-3.30)	-0.28 (-0.50)
TranspXauction	0.32 (0.79)	-0.28 (-0.75)
IlliquidXauction	1.75* (2.39)	-1.16 (-0.70)
ExposureXSP500	-3.81** (-3.99)	-1.96* (-2.47)
Chg_Default	2.18 (1.51)	2.34 (1.53)
<i>N</i>	1622	1622
<i>R</i> ²	0.195	0.175

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A8: Euro area aggregated into one country

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-2.91*	-2.62**	-3.25**	-0.69**	-0.01	-0.69**	-0.71**
	(-2.17)	(-5.03)	(-2.99)	(-3.81)	(-1.39)	(-3.38)	(-3.49)
TranspXauction	-0.70*	-0.71*	-0.64*	-0.64*	-0.64*	-0.87**	-0.87**
	(-2.12)	(-2.58)	(-2.09)	(-2.10)	(-2.06)	(-3.14)	(-3.16)
IlliquidXauction	1.68	0.59	0.74	0.91	0.60	0.86	0.91
	(1.07)	(1.01)	(1.14)	(1.55)	(0.75)	(1.24)	(1.29)
ExposureXSP500	-6.75**	-5.92**	-7.97**	-1.90**	-0.03**	-2.09**	-2.07**
	(-5.15)	(-4.65)	(-6.21)	(-21.48)	(-4.82)	(-16.40)	(-15.54)
Chg_Default	0.24	0.36	0.15	0.46	1.43	0.50	0.52
	(0.34)	(0.56)	(0.23)	(0.64)	(1.86)	(0.69)	(0.71)
<i>N</i>	3258	3258	3258	3258	2860	3258	3258
<i>R</i> ²	0.217	0.218	0.232	0.277	0.231	0.247	0.246

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A9: Changes in Exchange Rates added as independent variable

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
pct_fx	0.27 (0.83)	0.14 (0.45)	0.17 (0.54)	-0.10 (-0.32)	0.18 (0.58)	0.07 (0.24)	0.05 (0.18)
ExposureXauction	-2.71 (-1.95)	-2.50** (-4.83)	-3.03* (-2.67)	-0.64** (-3.95)	-0.01 (-1.31)	-0.65** (-3.48)	-0.68** (-3.61)
TranspXauction	-0.61* (-2.14)	-0.63** (-2.71)	-0.57* (-2.15)	-0.58* (-2.37)	-0.57* (-2.25)	-0.75** (-3.23)	-0.75** (-3.24)
IlliquidXauction	1.48 (0.91)	0.68 (1.23)	0.80 (1.19)	0.90 (1.68)	0.48 (0.61)	0.88 (1.42)	0.93 (1.50)
ExposureXSP500	-6.95** (-5.22)	-5.99** (-4.49)	-8.11** (-6.15)	-1.87** (-18.51)	-0.03** (-4.84)	-2.00** (-14.38)	-1.99** (-13.57)
Chg_Default	0.36 (0.52)	0.49 (0.76)	0.29 (0.43)	0.60 (0.84)	1.52* (2.08)	0.66 (0.90)	0.67 (0.92)
<i>N</i>	3991	3991	3991	3991	3600	3991	3991
<i>R</i> ²	0.202	0.200	0.213	0.241	0.209	0.217	0.217

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A10: Changes in Exchange Rates as dependent variable

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-0.13 (-1.62)	-0.05 (-0.65)	-0.10 (-0.85)	-0.03 (-1.11)	-0.00 (-1.64)	-0.03 (-1.28)	-0.04 (-1.49)
TranspXauction	0.02 (0.59)	0.01 (0.37)	0.02 (0.52)	0.02 (0.61)	0.05 (1.18)	0.02 (0.49)	0.02 (0.54)
IlliquidXauction	-0.00* (-2.02)	-0.00** (-2.71)	-0.00 (-1.98)	-0.00* (-2.65)	-0.00 (-0.15)	-0.00** (-2.66)	-0.00* (-2.38)
ExposureXSP500	-0.11 (-1.25)	-0.29** (-2.86)	-0.28* (-2.03)	-0.11** (-3.87)	-0.00 (-1.25)	-0.09** (-2.75)	-0.09* (-2.15)
Chg_Default	-0.01 (-0.17)	-0.02 (-0.32)	-0.02 (-0.37)	-0.02 (-0.42)	-0.05 (-0.64)	-0.01 (-0.28)	-0.01 (-0.25)
<i>N</i>	5778	5778	5778	5778	4897	5778	5778
<i>R</i> ²	0.012	0.022	0.017	0.035	0.015	0.022	0.021

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A11: LIBOR-OIS Spreads as dependent variable

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-0.61 (-1.49)	-0.03 (-0.67)	-0.14 (-1.00)	-0.10* (-2.51)	-0.00 (-1.50)	-0.06* (-2.33)	-0.07 (-1.73)
TranspXauction	0.03 (1.12)	-0.01 (-0.34)	0.00 (0.11)	0.01 (0.42)	0.02 (0.40)	-0.01 (-0.20)	0.00 (0.03)
IlliquidXauction	6.55 (1.59)	3.90 (1.32)	4.32 (1.44)	6.59* (2.25)	4.43 (2.00)	4.95 (1.77)	5.15 (1.84)
ExposureXSP500	-0.04 (-0.96)	-0.01* (-2.36)	-0.02 (-1.82)	-0.03* (-2.56)	-0.00* (-2.27)	-0.04* (-2.47)	-0.03* (-2.31)
Chg_Default	0.12 (1.73)	0.12 (1.70)	0.12 (1.71)	0.12 (1.68)	0.12 (1.68)	0.12 (1.67)	0.12 (1.68)
<i>N</i>	1154	1154	1154	1154	1154	1154	1154
<i>R</i> ²	0.072	0.067	0.067	0.077	0.068	0.074	0.075

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A12: Illiquidity proxied by ST Debt/GDP

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-2.66*	-2.57**	-3.19**	-0.63**	-0.01	-0.64**	-0.66**
	(-2.42)	(-4.90)	(-2.88)	(-3.96)	(-1.45)	(-3.55)	(-3.66)
Transp*auction	-0.57*	-0.59*	-0.52*	-0.56*	-0.53*	-0.72**	-0.72**
	(-2.08)	(-2.62)	(-2.06)	(-2.39)	(-2.14)	(-3.24)	(-3.25)
ill_gdp*auction	0.60	0.33	0.57	0.35	0.32	0.26	0.23
	(1.16)	(1.20)	(1.46)	(1.14)	(0.91)	(1.00)	(0.88)
Exposure*SP500	-6.96**	-6.03**	-8.14**	-1.85**	-0.03**	-1.99**	-1.98**
	(-5.16)	(-4.58)	(-6.18)	(-20.98)	(-4.86)	(-15.33)	(-14.56)
Chg_Default	0.36	0.49	0.29	0.61	1.52*	0.66	0.68
	(0.52)	(0.77)	(0.43)	(0.86)	(2.12)	(0.91)	(0.93)
Num of Obs	3953	3953	3953	3953	3555	3953	3953
R^2	0.200	0.200	0.212	0.240	0.209	0.216	0.216

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A13: Illiquidity proxied by ST Debt/Reserves

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-2.60*	-2.55**	-3.14**	-0.62**	-0.01	-0.64**	-0.66**
	(-2.51)	(-4.94)	(-2.92)	(-3.91)	(-1.46)	(-3.52)	(-3.63)
Transp*auction	-0.58*	-0.59*	-0.53*	-0.57*	-0.54*	-0.73**	-0.73**
	(-2.12)	(-2.66)	(-2.09)	(-2.42)	(-2.18)	(-3.27)	(-3.28)
ill_res*auction	0.09	0.02	0.09	0.00	0.04	-0.00	-0.01
	(0.85)	(0.22)	(0.95)	(0.02)	(0.44)	(-0.02)	(-0.12)
Exposure*SP500	-6.98**	-6.05**	-8.16**	-1.85**	-0.03**	-2.00**	-1.98**
	(-5.19)	(-4.58)	(-6.20)	(-21.05)	(-4.87)	(-15.37)	(-14.61)
Chg_Default	0.36	0.48	0.28	0.61	1.51*	0.66	0.68
	(0.52)	(0.76)	(0.43)	(0.85)	(2.11)	(0.91)	(0.93)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.200	0.200	0.212	0.240	0.209	0.217	0.216

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A14: Illiquidity proxied by Greenspan-Guidotti measure

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-2.18*	-2.09**	-2.49*	-0.44*	-0.01	-0.35	-0.39
	(-2.13)	(-4.04)	(-2.23)	(-2.16)	(-1.03)	(-1.19)	(-1.40)
Transp*auction	-0.18	-0.24	-0.20	-0.26	-0.18	-0.20	-0.24
	(-0.57)	(-0.74)	(-0.65)	(-0.72)	(-0.51)	(-0.43)	(-0.54)
ill_GG*auction	0.32	0.30	0.29	0.28	0.31	0.40	0.37
	(1.75)	(1.51)	(1.51)	(1.14)	(1.46)	(1.27)	(1.23)
Exposure*SP500	-6.96**	-6.04**	-8.16**	-1.85**	-0.03**	-2.00**	-1.98**
	(-5.19)	(-4.58)	(-6.20)	(-21.06)	(-4.87)	(-15.37)	(-14.61)
Chg_Default	0.36	0.48	0.29	0.61	1.50*	0.66	0.68
	(0.53)	(0.76)	(0.44)	(0.85)	(2.10)	(0.91)	(0.93)
Num of Obs	4005	4005	4005	4005	3607	4005	4005
R^2	0.201	0.200	0.212	0.240	0.209	0.217	0.216

t statistics in parentheses

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$

Table A15: Announcement Dates: Full Specification

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
ExposureXauction	-0.42 (-0.43)	-0.25 (-0.68)	-0.06 (-0.07)	0.16 (1.13)	-0.00 (-0.43)	0.14 (0.78)	0.13 (0.72)
TranspXauction	0.20 (0.70)	0.13 (0.50)	0.05 (0.17)	-0.05 (-0.23)	0.13 (0.47)	0.19 (0.76)	0.20 (0.81)
IlliquidXauction	-0.01 (-0.58)	0.02 (1.93)	0.01 (1.05)	0.00 (0.10)	0.01 (1.09)	-0.00 (-0.59)	-0.00 (-0.50)
ExposureXdate11	-14.76 (-1.62)	5.18 (0.66)	-3.48 (-0.37)	-2.69 (-0.87)	0.01 (0.25)	-0.56 (-0.19)	-0.91 (-0.30)
TranspXdate11	18.21** (3.33)	14.85** (2.85)	15.46** (2.89)	17.52** (3.30)	14.30* (2.56)	17.83** (3.47)	18.14** (3.49)
IlliquidXdate11	-0.33 (-1.52)	-0.22* (-2.02)	-0.23 (-1.91)	-0.28 (-1.49)	-0.31** (-3.05)	-0.42* (-2.60)	-0.40* (-2.39)
ExposureXdate12	-3.40 (-0.10)	-25.09 (-1.36)	-18.37 (-0.61)	-8.66 (-1.74)	-0.12 (-1.06)	-7.65 (-1.29)	-8.33 (-1.41)
TranspXdate12	-14.05* (-2.34)	-11.44* (-2.12)	-11.02 (-1.92)	-6.54 (-0.99)	-9.07 (-1.40)	-10.62 (-1.78)	-10.72 (-1.81)
IlliquidXdate12	-0.50 (-0.91)	-0.55 (-1.76)	-0.51 (-1.62)	-0.24 (-0.59)	-0.36 (-0.94)	-0.20 (-0.43)	-0.19 (-0.39)
ExposureXdate13	-42.89 (-1.24)	-39.94 (-1.15)	-42.76 (-1.26)	-7.40 (-0.75)	-0.39* (-2.07)	5.79 (0.58)	4.94 (0.49)
TranspXdate13	97.17** (4.99)	96.36** (5.02)	97.26** (4.97)	96.73** (4.83)	108.09** (5.21)	90.89** (5.09)	91.06** (5.11)
IlliquidXdate13	0.88 (0.26)	0.61 (0.18)	0.65 (0.19)	0.65 (0.19)	1.06 (0.31)	0.06 (0.02)	0.11 (0.03)
ExposureXdate14	-73.09 (-1.62)	-74.99** (-3.54)	-94.30** (-2.85)	-44.46** (-8.11)	-0.29 (-1.10)	-46.21** (-6.79)	-46.03** (-6.69)
TranspXdate14	-71.79** (-6.44)	-69.06** (-6.18)	-63.61** (-5.94)	-50.05** (-7.43)	-60.96** (-4.23)	-75.93** (-8.54)	-76.98** (-8.72)
IlliquidXdate14	0.68 (0.52)	-1.34* (-2.09)	-0.85 (-1.24)	0.34 (0.52)	-0.75 (-0.86)	0.81 (0.85)	0.72 (0.69)
ExposureXdate2	-85.29** (-4.50)	-80.45** (-5.04)	-102.39** (-5.56)	-14.05* (-2.22)	-0.05 (-0.13)	-17.45* (-2.51)	-17.99* (-2.64)
TranspXdate2	-25.29** (-3.15)	-24.10** (-3.13)	-21.22** (-2.70)	-31.06** (-3.84)	-30.52* (-2.57)	-37.98** (-4.89)	-38.17** (-4.92)
IlliquidXdate2	0.65 (1.76)	-0.52 (-1.65)	-0.26 (-0.60)	-0.36 (-1.30)	-0.72 (-1.08)	-0.17 (-0.47)	-0.17 (-0.46)
ExposureXSP500	-4.78** (-4.74)	-4.17** (-4.88)	-5.66** (-5.72)	-1.21** (-14.65)	-0.02** (-4.50)	-1.29** (-12.32)	-1.28** (-11.08)
Chg_Default	-0.05 (-0.15)	0.03 (0.08)	-0.11 (-0.34)	0.11 (0.28)	0.69 (1.73)	0.10 (0.28)	0.11 (0.30)
N	4000	4000	4000	4000	3601	4000	4000
R ²	0.309	0.313	0.319	0.356	0.318	0.343	0.343

OLS with country fixed effects, monthly time dummies, and robust standard errors.
t statistic in parentheses; * $p < 0.05$, ** $p < 0.01$

date11 refers to Dec 12, 2007. Lines introduced with ECB and SNB
date12 refers to Sep 18, 2008. Lines introduced with BOJ, BOE, and Bank of Canada
date13 refers to Sep 24, 2008. Lines introduced with Australia, Sweden, Denmark, and Norway
date14 refers to Oct 28, 2008. Lines introduced with NZL, Brazil, Mexico, Korea, and Singapore
date2 refers to Oct 13, 2008. Unlimited swaps introduced with ECB, BOE, BOJ, and SNB

Table A16: Non-OECD countries only

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Exports	Imports	Trade	Assets(TIC)	Assets(CPIS)	Debt	LTdebt
Exposure*auction	-3.11*	-2.66**	-2.64	-0.64*	-0.00	-0.68**	-0.72**
	(-2.21)	(-3.34)	(-1.91)	(-2.70)	(-0.52)	(-2.89)	(-3.00)
Transp*auction	-0.81	-1.06**	-0.96*	-0.93*	-1.18*	-1.19**	-1.19**
	(-1.86)	(-3.31)	(-2.47)	(-2.18)	(-2.66)	(-3.69)	(-3.69)
Illiquid*auction	2.48	1.27	1.08	1.56	1.03	1.85*	1.93*
	(1.63)	(1.74)	(1.44)	(1.92)	(1.19)	(2.40)	(2.46)
Exposure*SP500	-6.51**	-8.17**	-8.15**	-1.95**	-0.03**	-2.06**	-2.10**
	(-4.60)	(-5.63)	(-4.30)	(-20.52)	(-3.20)	(-14.82)	(-15.33)
Δ Default	-0.63*	-0.56*	-0.71**	-0.44	0.04	-0.30	-0.30
	(-2.47)	(-2.63)	(-3.04)	(-1.77)	(0.07)	(-1.02)	(-0.98)
Num of Obs	1785	1785	1785	1785	1387	1785	1785
R^2	0.283	0.282	0.289	0.389	0.283	0.333	0.337

t statistic in parentheses; OLS with country fixed effects, monthly time dummies, and robust standard errors.

* $p < 0.05$, ** $p < 0.01$